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RESPONSE TO COMMISSIONER'S INTERROGATORIES TO PARTIES' EXPERTS

State of Washington v. Allen Eugene Gregory No. 88086-7

KATHERINE BECKETT, PH.D, AND HEATHER EVANS, PH.D. UNIVERSITY OF WASHINGTON

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Interrogatory 1

Are you aware of cases that are in fact missing from the trial reports, or should your statement be taken as a caveat that you have not independently verified the inclusiveness of the trial reports?

We would like to address the first question even though it was directed to Dr. Scurich in order to provide relevant context. The data set for the study consisted of all Trial Reports (TRs) filed with the Supreme Court through May of 2014 (numbers 1-331). In November 2013, Mr. Gregory's attorneys filed a Motion to Complete the Process of Compiling a Full Set of Aggravated Murder Reports. This motion included attachments documenting over thirty aggravated murder cases for which final trial reports were missing. In January 2014, the Supreme Court denied the motion. The Court had already ruled that the data set was complete enough to perform proportionality review, the purpose of which "is to avoid random arbitrariness and imposition of the death sentence based on race."¹

Although the Supreme Court ruled that the data set was complete enough to perform proportionality review, several missing trial reports were filed between January and May of 2014, and we performed the study anew in order to use the most comprehensive data set possible.² Trial reports are still missing for some of the aggravated murder cases listed in the Motion to Complete. However, only three missing reports relate to special sentencing proceedings; all three of these proceedings involved white defendants who received life sentences.³

Interrogatory 2

Please provide a fuller description of the methods utilized by coders who coded mitigating circumstances, identify any applicable written protocols for this coding, and explain any unwritten data coding protocols that were used.

¹ In re Elmore, 162 Wn.2d 236, 270, 172 P.3d 335 (2007).

² In several places in our *Response to Evaluation*, we represented the total sample size as 297 cases. Because we included TR 34A in these analyses, the correct number is 298.

³ These three are: (1) Duane Bartholomew's second special sentencing proceeding following reversal of death sentence, remand, and new jury trial resulting in life without parole sentence (*see* TR 3); (2) Mitchell Rupe's third special sentencing proceeding following first two reversals and final jury trial resulting in life without parole sentence (*see* TR 57 & 31); and (3) Charles Finch's second special sentencing proceeding following reversal of death sentence, remand, and new jury trial resulting in life without parole sentence (*see* TR 154). Although a jury entered a life sentence in Mr. Finch's second trial, he committed suicide before the judgment was entered. Arguably, no trial report need be entered as no judgment was entered.

The Commissioner correctly identified the coding protocol for the mitigating circumstances. Specifically, the protocol was as follows:

- 1. Enter the number of discrete statutory mitigating circumstances from question 3(c) into the Mit_Circum_Statutory field.
 - a. If the "No" box is checked, enter zero.
 - b. If the "Yes" box is checked with no description, enter one.
 - c. If nothing is checked and the description field is empty, enter zero.
 - d. If nothing is checked but the judge has described one or more mitigating circumstances, count them and enter the appropriate number.
 - e. If one or more mitigating circumstances in this field is actually non-statutory, count it in the Mit_Circum_Non-Stat field instead.
- 2. Under question 3(d), evaluate the number of individual concepts described by the trial judge and enter that number into the Mit_Circum_Non-Stat field.
 - a. If the "No" box is checked, enter zero.
 - b. If the "Yes" box is checked with no description, enter one.
 - c. If nothing is checked and the description field is empty, enter zero.
 - d. If nothing is checked but the judge has described one or more mitigating circumstances, count them and enter the appropriate number.
 - e. If a one or more descriptions in this field is actually statutory, count it in the Mit_Circum_Statutory field instead.
- 3. Add the numbers from 3(c) and 3(d) and enter the total into the TotMitCircum field.

Where a factor could fairly be characterized as either statutory or non-statutory, it was left in its original category. Also, please note that steps (1)(e) and (2)(e) above were not critical because the total number of mitigating circumstances was used in the analyses.

The Commissioner correctly identified one data entry error: The number of mitigating circumstances for TR 25 should have been 2, not 1. In light of this discovery, we re-checked *all* trial reports for cases that had a special sentencing proceeding, and confirmed that TR 25 was the only one with such an error. We did, however, discover that a 0 rather than a 1 had been entered for the number of aggravating circumstances found by the judge/jury for the proceeding associated with TR 34A, so also corrected this.

Correcting these isolated errors and re-running the analysis does not alter the regression results in a meaningful way. Specifically, for the model presented in Table 7 of the *Updated Report*, the p-value for Black defendant was 0.040; after the changes, this p-value is 0.039. The coefficient for this variable was previously reported as 1.573 (meaning that Black defendants were 4.82 times as likely than non-Black defendants to be sentenced to death); after the corrections, this figure changes very slightly to 1.582. When transformed to an odds ratio, the latter figure indicates that Black defendants were 4.86 times as likely as non-Black defendants to be sentenced to death. Table 1 below shows these results. The unaltered output associated with these analyses is shown beneath the table.⁴

Table 1. Revised Table 7 from Updated Report with Data Entry Errors Corrected and TR 34AAdded: Impact of Case Characteristics and Defendant Race on Capital Sentencing Outcomesin Death Eligible Cases, December 1981 - May 2014

N= 77	Death	Pseudo R ² = .2361		
Variable	Coefficient	Exact	Odds	Referent
		P-Value	Ratio	(Compared to)
Prior Convictions (logged)	-0.091	.510	.913	
1 Victim	-0.722	.221	.486	Multiple victims
Aggravating Circumstances	0.630	.016	1.88**	
Mitigating Circumstances (logged)	-0.258	.089	.773*	
Defenses	-0.794	.034	.452**	
Victim Held Hostage	0.717	.222	2.05	Not held hostage
Black Defendant	1.582	.039	4.86**	Non-black
* significant at α = .10 ** significan	t at α = .05	***	' significant	at α = .01

UNALTERED STATISTICAL OUTPUT ASSOCIATED WITH TABLE 1: DATA ENTRY ERRORS CORRECTED AND TR 34A ADDED logit DP_Sentence lnPriors Vics_1Total AppliedAggCir_Num LnTotMitCircum Defenses_Num Vics AnyHostage D RaceB , level(90) ;

Iteration 0: log likelihood = -52.583924Iteration 1: log likelihood = -40.664023 Iteration 2: log likelihood = -40.170014 Iteration 3: log likelihood = -40.166274 Iteration 4: log likelihood = -40.166273 Number of obs = 77 LR chi2(7) = 24.84 Prob > chi2 = 0.0008 Logistic regression Log likelihood = -40.166273Pseudo R2 = 0.2361 _____ DP Sentence | Coef. Std. Err. z P>|z| [90% Conf. Interval] _____ lnPriors | -.0913328 .1385858 -0.66 0.510 -.3192863 .1366206 Vics 1Total | -.7215931 .5896597 -1.22 0.221 -1.691497 .2483108
 AppliedAggCir_Num |
 .6299845
 .2624197
 2.40
 0.016
 .1983425
 1.061627
 LnTotMitCircum | -.2575945 .1513135 -1.70 0.089 -.506483 -.0087059 Defenses Num | -.7935932 .3740099 -2.12 0.034 -1.408785 -.1784016 Vics_AnyHostage | .7169782 .5866349 1.22 0.222 -.2479504 1.681907 D RaceB | 1.581795 .767834 2.06 0.039 .3188205 2.84477 cons | -1.114036 .7731921 -1.44 0.150 -2.385824 .1577521

⁴ Throughout this document, the output shown is unaltered other than the fact that significant findings have been bolded.

To be clear, these results show that Black defendants are 4.86 times as likely as other defendants to be sentenced to death after controlling for the other, legally relevant case characteristics included in the model. To put the magnitude of this effect in context, this means that the substantive impact of a defendant being Black on sentencing outcomes is greater than the impact of having four additional aggravating circumstances.⁵ This finding regarding defendant race is significant at p=.039.

Please note that when re-analyzing data in response to subsequent interrogatories, we always use the updated data set that corrects all data entry errors.

Interrogatories 4-6

Are the following trial reports the reports that relate to the denominator of 14 black defendants in Table A of the *Response to Evaluation* at 16: Trial Reports 29, 77, 88, 119, 135, 157, 177, 180, 185, 186, 194, 216, 281, 312? If your response is "no," please list the trial reports that relate to this denominator.

Are the following trial reports the reports that relate to the denominator of 57 white defendants in Table A of the *Response to Evaluation* at 16: Trial Reports 2,3,7, 9, 15, 20, 23, 25, 26, 31, 34, 34a, 36, 39, 42, 43, 44, 45, 47, 48, 51, 52, 53, 56, 58, 62, 63, 64, 65, 66, 75, 76, 86, 92, 93, 95, 125, 132, 140, 144, 154, 164, 165, 167, 174, 175, 176, 182, 183, 184, 190, 220, 227, 251, 258, 303, 313? If your response is "no," please list the trial reports that relate to this denominator.

Are the following trial reports the reports that relate to the denominator of 1 "other race" defendants in Table A of the *Response to Evaluation* at 16: Trial Reports 8, 13, 14, 60, 158, 160, 181, 197, 224, and 256? If your response is "no," please list the trial reports that relate to this denominator.

In each of these interrogatories we were asked to confirm that the listed Trial Reports correspond to a specific group of defendants in our *Response to Evaluation*. In each case, we confirmed that this was the case. The answer to these questions is yes.

Interrogatories 7 and 8

Please indicate whether Trial Reports 92, 167, 182, and 224 were in fact included in the set of cases used to calculate the percentages of aggravated murder cases with special sentencing proceedings in which juries imposed a death sentence, by race of defendant. If these trial reports were included in this set of cases, please indicate whether you maintain they were properly included and, if so, the basis for this position.

⁵ An odds ratio of 4.86 means that Black defendants are 386% more likely than non-Black defendants to be sentenced to death. An odds of ratio of 1.88 means that each additional aggravating circumstance increases the likelihood that a death sentence will be imposed by 88%. 386% divided by 88% = 4.4.

Please indicate whether cases that are the subject of Trial Reports 92, 167, 182, and 224 were or were not included in the regression analysis relating to special sentencing proceedings in which juries imposed a death sentence. If any of the cases that are the subject of these trial reports were included in the regression analyses, please explain the reasons for such inclusion.

Before answering the central question, we would like to clarify two issues. First TRs 92, 167, 182 and 224 *were* included in all of the regression analyses as well as in the descriptive analyses. By contrast, TRs 81, 152, and 153 were included in the descriptive and regression analyses of prosecutorial *filing* decisions, but excluded from the descriptive and regression analyses of *sentencing* decisions.

The principle that guided this decision is as follows: because our second set of analyses assesses whether race and other factors affect sentencing outcomes, we included all special sentencing proceedings regardless of whether a judge or jury served as the decision-maker in the proceeding. As we stated in our *Updated Report*, in most cases a jury served as the decision-maker, but in some, the defendant waived his or her right to a jury and the judge served as the decision-maker (see footnote 5, p. 19).

We included all special sentencing proceedings in our analyses of sentencing decisions because if a special sentencing proceeding occurred, a sentence of death *could* have been imposed by the judge or jury. The difference between the included and excluded cases is that in the former, a special sentencing proceeding actually occurred. Although it is true that in three of these four cases the prosecution stipulated during this proceeding that it could not prove beyond a reasonable doubt that there were not sufficient mitigating circumstances to warrant leniency, the death notices do not appear to have been withdrawn, special sentencing proceedings actually took place, and the judge, exercising independent judgment based upon the facts of the crime, could still have imposed a death sentence.

By contrast, in the excluded cases (TRs 81, 152 and 153), there was no evidence that a special sentencing proceeding actually occurred. Although the judge did enter a date of a special sentencing proceeding, it is clear that this was simply the date on which the sentence was imposed; no information about a special sentencing proceeding is provided. Instead, these trial reports indicate that there was a guilty plea with an agreement of a life without parole sentence, suggesting that the death notice had been withdrawn. It thus appears that in these cases, the defendant was sentenced for non-capital aggravated murder after offering a guilty plea based on an agreement of a life without parole sentence. In these cases, the judge was precluded from sentencing the defendant to death as a result of the structure of the plea deal. The judge would not have had the legal ability to weigh the facts of the crime and determine

whether death was appropriate. For this reason, these three cases were not included in our analyses of sentencing outcomes.⁶

Interrogatory 10

Do you maintain the accuracy of the description of Table 1 as showing the share of convictions that resulted in a death sentence? If yes, please explain.

Table 1 of our *Updated Report* shows the proportion of aggravated murder cases with deatheligible defendants in which a death sentence was sought and imposed, by county. As the title of this table and the surrounding discussion suggests, the unit of analysis is case, not conviction. We inadvertently used the word "conviction" in the penultimate sentence of p. 14 of the *Response to Evaluation* when discussing these findings; we should have used the word "case" or "proceeding" in this sentence. The title of the table and all other references to the table correctly identify the unit of analysis as "case," not conviction. The numbers shown in the table are correct.

Please note: We recognize that the term "case" has a specific legal meaning. However, in our analyses of sentencing outcomes, we use the word "case" to refer to our data points, namely, special sentencing proceedings.

Interrogatory 11

Is it correct that you added the "unknown" case to the numerator but mistakenly did not add it to the denominator? If no, please explain the basis for the numerator and denominator.

The Commissioner is correct: we added the case (i.e. proceeding) in which the race of defendant is unknown to the numerators but not the denominators in Table 2 of our *Updated Report*. Including this case in the denominator does not alter the percentages reported in the table. In the "death notice filed" column, adding this case to the denominator leads to a result of 28.96 percent, which would have been rounded to 29 percent, as is shown. In the "death penalty imposed" column, adding this proceeding to the denominator leads to a result of 11.78 percent, which would have been rounded to 12 percent, as is shown. Finally, adding this proceeding to the denominator leads the result to 4.38 percent, which would have been rounded to 4 percent, as is shown.

This proceeding was also excluded from the figures presented in the bottom row of Table B of our *Response to Evaluation*, but again, including it does not meaningfully alter the results. Below, we first present Table B of the *Response to Evaluation*, then Table 2 which updates

⁶ These three proceedings all involved white defendants who killed multiple victims.

these figures to include the omitted case and to reflect the data entry corrections. The figures that differ between Table B and Table 2 are bolded.

by Race of Defendant (From <i>Response to Evaluation</i>)										
	Nun aggr circur	nber of avating nstances	Number of mitigating circumstances		ber of ating stances		of Number of violent prior convictions		Victim held hostage	Total number of cases
	Mn	Md	Mn	Md	Mn	Md	Mn	Md	Percent	N
Black Defendants	2	1	2	2	1	1	2	2	42.9% (6/14)	14
Non-Black Defendants	2	2	2	2	2	1	1	1	31.8% (21/66)	67
White Defendants	3	2	2	2	2	1	1	1	35.1% (20/57)	57
Other Defendants	2	2	2	1	4	1	1	0	11.1% (1/9)	10
All Defendants	2	2	2	2	2	1	1	1	33.3% (27/81)	81

 All
 2
 2
 2
 2
 2
 2
 1
 1
 33.3% (27/81)
 81

 Notes: Mn represents mean (average); Md represents median (the typical value). Defendant race is unknown in one case; the number of defendants broken out by race is 81 although the total number of all defendants with death notices filed and special sentencing proceedings is 82. Here we refer to the number of aggravating circumstances found by the jury (as opposed to the number alleged by prosecutors). Information on whether

 Table 2. Comparison of Case Characteristics in Cases with Special Sentencing Proceedings,

 by Race of Defendant, Including Case with Racially Unidentified Defendant

the victim was held hostage was missing in one Trial Report (n=81).

by have or berendant, melading case with hacidity officient berendant										
	Num aggra circum	ber of wating stances	Number of mitigating circumstances		Number of violent prior convictions		Victim held hostage	Total number of cases		
	Mn	Md	Mn	Md	Mn	Md	Mn	Md	Percent	N
Black Defendants	2	1	2	2	1	1	2	2	42.9% (6/14)	14
Non-Black Defendants	2	2	2	2	2	1	1	1	31.3% (21/67)	67
White Defendants	3	2	3	2	2	1	1	1	35.1% (20/57)	57
Other Defendants	2	2	2	1	4	1	1	0	11.1% (1/9)	10
All Defendants	2	2	2	2	2	1	1	1	33.3% (27/81)	82

Notes: See Table B notes above.

Interrogatory 13

Is the difference in the 85 cases used in the *Updated Report* at 21 and the 81 cases that comprise the denominators used in Table A in the Response to Evaluation at 16 due to the removal of Trial Reports 68, 152, 153, 217, and 308 and the addition of Trial Report 34A? If not, please explain.

Yes, the removal of TRs 68, 152, 153, 217 and 308, and the addition of TR 34A, explains the difference in the denominators in these documents.

Interrogatory 14

Please indicate whether you agree or disagree with the identification of the cases that are appropriately included in the calculations of the percentages. If you agree, please recalculate the resulting percentages. If you do not agree, please explain.

Yes, the cases that were included in Table A should have been included. However, we should not have specified in the title that the death sentence must have been imposed by a jury; it could have been imposed by either a judge or jury. The correct title of Table A is: *Percent of Special Sentencing Proceedings in which a Death Sentence was Imposed, by Race of Defendant.*

Interrogatory 15

Please list the trial report numbers of cases included in the "death penalty retained" column of Table 2 and further explain why the listed cases are considered "retained."

We included in the "retained" category any defendant sentenced to death under the current statute whose death sentence has not, to date, been reversed. This includes people who have been executed or who remain on death row. The list of such individuals is available through the Washington State Department of Corrections (DOC).⁷

At the time we accessed this information, a total of 13 individuals met these criteria. These included five people who have been executed: Cal Brown (TR 140, white); Charles Campbell (TR 9, white); Westley Dodd (TR 76, white); James Elledge (TR 183, white); and Jeremy Sagastegui (TR 160, other). They also included eight people who remain on death row: Dayva Cross (TR 220, white); Cecil Davis (TR 180, black); Clark Elmore (TR 165, white); Jonathan Gentry (TR 119, black); Allen Gregory (TR 216, black); Conner Schierman (TR 303, white); Byron Scherf (TR 313, white); Dwayne Woods (TR 177, black); and Robert Yates (TR 251, white). Because Mr. Woods died of a cardiac arrest while on death row in January of 2017, he no longer appears on the DOC list of people currently on death row.

⁷ See <u>http://www.doc.wa.gov/docs/publications/reports/100-SR001.pdf</u> and <u>http://www.doc.wa.gov/docs/publications/reports/100-SR002.pdf</u>

Interrogatory 17

Please list the trial report numbers of cases included in the "death penalty retained" column of Table 3 and further explain why the listed cases are considered "retained."

We included in the "retained" category any defendant sentenced to death under the current statute whose death sentence has not, to date, been reversed (see our answer to Interrogatory 15). In Table 3 of our *Updated Report*, we included only cases that involved a single white or black victim, as described in the note that appears beneath the table. Of the 13 cases in the general retained category, we included five that involved a single black or white victim. These are: Elmore (TR 165), Gentry (TR 119), Gregory (TR 216), Elledge (TR 183) and Brown (TR 140). Scherf (TR 313) was incorrectly coded as not having the death penalty retained so was excluded from this table.

Adding TR 313 has no impact on the percentages shown in the table. Specifically, if we add this case (which involved a white defendant and a single white victim) to the figures reported in Table 3 of the *Updated Report*, we can calculate that the percentage of cases involving a white defendant and a single white victim for which the death penalty has been retained remains 3 (3.4) percent. It is also worth noting that "retained" status is not included in any of the regression analyses, so the regression results are also unaffected by the omission of Scherf (TR 313) from Table 3 of the *Updated Report*.

Interrogatory 19

What p-value results if the model reported in Table 7 of the *Updated Report* is run with the three coding errors corrected, the first sentencing proceedings reported in Trial Reports 7, 180, and 216 removed, and with the logarithmic transformations of variables set forth in the Response to Evaluation?

In these interrogatories, we are asked to report the p-values that result if the model described in Table 7 of the *Updated Report* is run with the three data entry errors corrected, TRs 7, 180 and 216 removed, and with the logarithmic transformation of variables described in our *Response to Evaluation*. In Interrogatory 20, we are asked to also include the case associated with TR 34A in this model.

To be clear, the results presented in our *Response to Evaluation* already reflect the correction of the three previously identified data entry errors, the logarithmic transformation of variables, and the proceeding associated with Trial Report 34A. It is our practice to always re-analyze data with corrections when data entry errors have been identified. It is also our practice to always perform appropriate transformations of variables.

We believe that excluding the proceedings described in TRs 7, 180 and 216, as suggested by Dr. Scurich, is highly improper for the reasons we articulated in our *Response to Evaluation* and are described below. Our analyses of sentencing outcomes are designed to assess whether race and other factors influence sentencing decisions in Washington State capital cases in which a death sentence may be imposed. The unit of analysis in these models is the sentencing proceeding, not the defendant. That is, each data point in the analysis represents a proceeding and associated decision – in this case, the decision to impose a sentence of death.

Intentionally excluding proceedings in which a death sentence could have been imposed from the analysis is improper, as the second trials of the defendants involved in TRs 7, 180 and 216 occurred substantially later than the first, involved different juries and, in two of the three instances, different case characteristics. As a result of these meaningful differences, the first and second proceedings could very well have resulted in a different sentencing outcome. In Mr. Gregory's case, for example, the number of mitigating circumstances and the number of prior convictions were different in the two proceedings. In addition, Mr. Gregory's special sentencing proceedings were separated by eleven years and involved different juries and defense attorneys (see TRs 216, 312). The characteristics associated with Mr. Davis' first and second special sentencing proceedings were also quite different: the two trials were separated by nine years, involved different juries and defense attorneys, and the number of prior convictions was different in the two proceedings (see TRs 180, 281).

Moreover, there is no rational basis for deciding which of a defendant's two special sentencing proceedings to include. Deciding to keep the first but exclude the second, or vice versa, would be arbitrary, but because the proceedings involve different characteristics, either decision could have an impact on the results.

Dr. Scurich suggested that including both the first and second cases of the three defendants who had two special sentencing proceedings violates the assumption that the cases included in the regression model are independent (see pp. 25-27 of his critique). Although it is true that a defendant's second special sentencing proceeding is not entirely independent of his or her first proceeding, it is also true that any proceedings adjudicated by the same judge, involving the same attorneys, or adjudicated in the same county also violate the assumption of independence. Nonetheless, as a practical matter, researchers using regression methods to analyze sentencing outcomes routinely include cases that involve the same judges, attorneys,

counties, and defendants in their analyses, and the results are often published in well-regarded, peer-reviewed journals.⁸

For all of these reasons, we strongly believe that it is highly inappropriate to remove these three proceedings from the analysis, just as it would be improper to exclude second proceedings (TRs 31, 281 and 312) and include only the first proceedings. It is also inappropriate to exclude the proceeding associated with TR 34A, which was included in the analyses presented in our *Response to Evaluation*, because TR 34A represents a completely different crime and jury proceeding than TR 34. Below, we exclude these cases despite these concerns in order to be responsive to the Commissioner's request. For reference, Table 1 of this document is copied below, which presents the results of the model that includes all relevant cases, including Trial Report 34A, the appropriate logarithmic transformation of variables is performed, and data entry errors have been corrected.

N= 77	Death Penalty Imposed Pseudo R ² = .								
Variable	Coefficient	Exact	Odds	Referent					
		P-Value	Ratio	(Compared to)					
Prior Convictions (logged)	-0.091	.510	.913						
1 Victim	-0.722	.221	.486	Multiple victims					
Aggravating Circumstances	0.630	.016	1.88**						
Mitigating Circumstances (logged)	-0.258	.089	.773*						
Defenses	-0.794	.034	.452**						
Victim Held Hostage	0.717	.222	2.05	Not held hostage					
Black Defendant	1.582	.039	4.86**	Non-black					
* significant at α = .10 ** significan	t at α = .05	***	' significant	at α = .01					

Table 1. Revised Table 7 from *Updated Report* with Data Entry Errors Corrected and TR 34A Added: Impact of Case Characteristics and Defendant Race on Capital Sentencing Outcomes in Death Eligible Cases, December 1981 - May 2014

⁸ For examples in which Washington State sentencing data are analyzed, and the sentencing outcome rather than the defendant is the unit of analysis, see Randy R. Gainey, Sara Steen and Rodney L. Engen, "Exercising Options: An Assessment of the Use of Alternative Sanctions for Drug Offenders", *Justice Quarterly* 22:4, 488-520 (2005); Rodney L. Engen, Randy R. Gainey, Robert D. Crutchfield, and Joseph G. Weis, "Discretion and Disparity Under Sentencing Guidelines," *Criminology* 41, 1: 99-130 (2003); Rodney L. Engen, "The Power to Punish: Discretion and Sentencing Reform in the War on Drugs," *American Journal of Sociology* 105, 5: 1357-1395 (2000); and Alexes Harris, Heather Evans and Katherine Beckett, "Courtesy Stigma and Monetary Sanctions: Toward a Socio-Cultural Theory of Punishment," *American Sociological Review* 76, 2: 234-64 (2011).

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The Commissioner requests that we present this model without Trial Report 34A and removing three other relevant cases: Trial Reports 7, 180 and 216. The results are shown in Table 3 below. The unaltered statistical output appears beneath the table.

Table 3. Re-Run of Table D from Response to Evaluation (see notes for details)

Impact of Case Characteristics and Defendant Race on Capital Sentencing Outcomes in Cases with Special Sentencing Proceedings, December 1981 - May 2014

Deat	h Penalty Impos	Pseudo R ² = 0.2142	
			LR chi2(7) = 21.18
			Prob > chi2 = 0.0035
Coefficient	Exact	Odds	90% Confidence Interval
	P-Value	Ratio	
-0.040	0.787	0.961	284, .204
-0.653	0.266	0.520	-1.62, .313
0.568	0.028	1.764**	.144, .991
-0.257	0.103	0.773	517, .003
-0.706	0.059	0.494*	-1.32,091
0.739	0.209	2.094	228, 1.71
1.392	0.075	4.022*	.104, 2.68
	Deat Coefficient -0.040 -0.653 0.568 -0.257 -0.706 0.739 1.392	Death Penalty Impos Coefficient Exact P-Value -0.040 0.787 -0.653 0.266 0.568 0.028 -0.257 0.103 -0.706 0.059 0.739 0.209 1.392 0.075	Death Penalty Imposed Coefficient Exact Odds P-Value Ratio -0.040 0.787 0.961 -0.653 0.266 0.520 0.568 0.028 1.764** -0.257 0.103 0.773 -0.706 0.059 0.494* 0.739 0.209 2.094 1.392 0.075 4.022*

* significant at $\alpha = .10$ ** significant at $\alpha = .05$ *** significant at $\alpha = .01$

Note: In this m	odel, Trial Reports	7, 180, 216 and 34A	were removed.
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UNALTERED STATISTICAL OUTPUT ASSOCIATED WITH TABLE 3							
logit DP_Sentence Defenses_Num Vics_	lnPriors Vics AnyHostage D_	a_1Total App RaceB, level	pliedAgg 1(90) ;	Cir_Num	LnTotMitCircu	m	
Iteration 0: log Iteration 1: log Iteration 2: log Iteration 3: log Iteration 4: log	likelihood = likelihood = likelihood = likelihood =	-49.436013 -39.188099 -38.849462 -38.847338 -38.847337					
Logistic regression Log likelihood = -38.847337				umber of R chi2(7) rob > chi seudo R2	obs = = 2 = 0 = 0	73 21.18 .0035 .2142	
DP_Sentence	Coef.	Std. Err.	Z	₽> z	[90% Conf.	Interval]	
InPriors Vics_1Total AppliedAggCir_Num LnTotMitCircum Defenses_Num Vics_AnyHostage D_RaceB cons	0401487 6533524 .5676415 257243 7058531 .7392167 1.391851 -1.135606	.1485276 .5876376 .2575915 .1579313 .3736433 .5882227 .7827078 .7708462	-0.27 -1.11 2.20 -1.63 -1.89 1.26 1.78 -1.47	0.787 0.266 0.028 0.103 0.059 0.209 0.075 0.141	2844549 -1.61993 .1439412 5170169 -1.320442 2283235 .1044113 -2.403535	.2041574 .3132254 .9913417 .0025309 0912646 1.706757 2.679291 .1323229	

Thus, when the three disputed cases and TR 34A are excluded, the results continue to indicate that the race of defendant has a substantial and statistically significant impact on sentencing outcomes. Specifically, removal of these cases reduces the coefficient for Black defendant modestly, from 1.58 to 1.39. When expressed as log odds, the latter indicates that Black defendants are 4.02 times as likely than non-Black defendants to be sentenced to death after controlling for the other variables in the model.

Removal of the disputed cases as well as TR 34A increases the p-value from .041 to .075, both of which indicate a statistically significant effect in the context of this study. As we have explained, the latter p-value (p= .075) does indicate a statistically significant relationship, as the appropriate alpha level in this study is .10 rather than .05. To reiterate, this is because the primary hypothesis being tested (i.e. that Black defendants are more likely to be sentenced to death than other, similarly situated defendants) is directional, and is therefore appropriately paired with an alpha level of .10.⁹ The directionality of this hypothesis is grounded in the empirical record. As the literature review presented in our *Updated Report* shows, when studies find evidence that race matters, they find that a) Black/minority defendants are treated comparatively harshly; and/or that b) defendants convicted of killing White victims are treated any studies in Washington State or the United States that show that White defendants or defendants convicted of killing people of color are treated comparatively harshly. As a result,

⁹ Pillemer, David, *One-versus Two-Tailed Hypothesis Tests in Contemporary Educational Research*, EDUCATIONAL RESEARCHER, 20, 9: 13-17 (1991); Ringwalt, C., Paschall, M. J., Gorman, D., Derzon, J., & Kinlaw, A., *The use of one-versus two-tailed tests to evaluate prevention programs*, EVALUATION & THE HEALTH PROFESSIONS 34, 2: 135-150 (2011); Agresti, A. and B. Finlay, *One-sided alternative hypotheses*, STATISTICAL METHODS FOR THE SOCIAL SCIENCES (Upper Saddle, NJ: Prentice Hall, 1997, 3rd edition) at 165-166.

¹⁰ See especially Jennifer L. Eberhardt, Paul G. Davies, Valerie J. Purdie-Vaughns & Sheri Lynn Johnson, *Looking Deathworthy: Perceived Stereotypicality of Black Defendants Predicts Capital-Sentencing Outcomes*, 17 PSYCHOLOGICAL SCIENCE 383 (2006); Phillip Atiba Goff, Jennifer L. Eberhardt, Melissa J. Williams & Matthew Christian Jackson, *Not Yet Human: Implicit Knowledge, Historical Dehumanization, and Contemporary Consequences*, 94 J. PERS. & SOC. PSYCHOL. 292 (2008); Radha Iyengar, *Who's the Fairest in the Land? Analysis of Judge and Jury Death Penalty Decisions*, 54 J. L. & ECON. 693, 695–96, 708 (2011); Justin D. Levinson, Robert J. Smith & Danielle M. Young, *Devaluing Death: An Empirical Study of Implicit Racial Bias on Jury-eligible Citizens in Six Death Penalty States*, 89 N.Y.U. L. REV. 513 (2014); Tara L. Mitchell, Ryann M. Haw, Jeffrey E. Pfeifer & Christian A. Meissner, *Racial Bias in Mock Juror Decision-Making: A Meta-Analytic Review of Defendant Treatment*, 29 LAW & HUMAN BEHAV. 621, 631 (2005); Mona Lynch & Craig Haney, *Looking Across the Empathic Divide: Racialized Decision Making on the Capital Jury*, 2011 MICH. ST. L. REV. 573 (2011); Mona Lynch & Craig Haney, *Emotion, Authority and Death: (Raced) Deliberations in Mock Capital Jury Deliberations*, 40 LAW & Soc. INQUIRY 377 (2015).

testing a directional hypothesis and the use of the p < .1 threshold is appropriate. Setting the alpha level at .10 is also standard practice in this area: studies of capital sentencing published in highly regarded and peer-reviewed journals include an alpha level of .10.¹¹ Because .075 is less than .10, this p-value indicates a statistically significant result.

To summarize, relevant cases should not be removed from the analysis. However, even when these four relevant trial reports are excluded, the findings continue to indicate that Black Defendants are four times as likely to receive the death penalty. The p-values also remain below the .10 threshold.

Interrogatory 20

What p-value results if the model reported in Table 7 of the *Updated Report* is run with the three coding errors corrected, the first sentencing proceedings reported in Trial Reports 7, 180, and 216 removed, Trial Report 34A added, and with the logarithmic transformations of variables set forth in the *Response to Evaluation*?

Please see our response to Interrogatory 19 regarding the propriety of excluding these three cases. We nevertheless comply with the Commissioner's request below.

When the three disputed cases are removed, TR 34A is included, and the appropriate logarithmic transformation of variables is conducted, the model results continue to indicate that the race of defendant has a substantial and statistically significant impact on sentencing outcomes (see Table 4 and associated output below). Specifically, removal of these three relevant cases reduces the coefficient for Black defendant modestly, from 1.58 to 1.44. When expressed as log odds, these results indicate Black defendants are 4.2 times as likely than non-Black defendants to be sentenced to death after controlling for the other variables in the model - despite the fact that two of the three excluded proceedings involved Black defendants who were sentenced to death. Removal of the disputed cases increases the p-value from .039 to .066, but both of these values indicate a statistically significant effect in the context of this study for the reasons described above.

¹¹ See, for example, David C. Baldus, Catherine M. Gross, George Woodworth and Richard Newell, *Racial Discrimination in the Administration of the Death Penalty: The Experience of the United States Armed Forces (1984-2005), JOURNAL OF CRIMINAL LAW & CRIMINOLOGY 101, 4: 1227-1336 (2012); John Donahue III, <i>Empirical Evaluation of the Connecticut Death Penalty System Since 1973: Are There Unlawful Racial, Gender and Geographic Disparities?* JOURNAL OF EMPIRICAL LEGAL STUDIES 11, 4: 637-96 (2014).

Table 4. Re-Run of Table D from *Response to Evaluation* (see notes for details)

Impact of Case Characteristics and Defendant Race on Capital Sentencing Outcomes in Cases with Special Sentencing Proceedings, December 1981 - May 2014

N - 74	Death Penalty Imposed Pseudo $R^2 = 0.215$						
N= /4	Deat	in renarcy impos	beu	1 Seudo IX = 0.2180			
				LR chi2(7) = 21.84			
				Prob > chi2 = 0.0027			
Variable	Coefficient	Exact	Odds	90% Confidence Interval			
		P-Value	Ratio				
Prior Convictions (In)	-0.041	0.772	0.960	287, .203			
1 Victim	-0.666	0.258	0.514	-1.63, .308			
Aggravating Circumstances	0.582	0.024	1.79**	.162, 1.01			
Mitigating Circumstances (In)	-0.240	0.118	0.787	503, .009			
Defenses	-0.747	0.044	0.474**	-1.34,125			
Victim Held Hostage	0.754	0.200	2.13	218, 1.72			
Black Defendant	1.44	0.066	4.21*	.142, 2.71			
* significant at α = .10	** significant at	α = .05	cant at α = .01				

Note: In this model, Trial Reports 7, 180, 216 were removed.

UNALTERED STATISTIC	CAL OUTPUT ASS	SOCIATED WITH	TABLE 4	1				
Iteration 4: log likelihood = -39.040982								
Logistic regression Log likelihood = -3	n 39.040982		Nu LI Pu Ps	umber of c R chi2(7) rob > chi2 seudo R2	bbs = = 2 = 0 = 0	74 21.84 .0027 .2186		
DP_Sentence	Coef.	Std. Err.	Z	P> z	[90% Conf.	Interval]		
lnPriors Vics_1Total AppliedAggCir_Num LnTotMitCircum	<pre>0412258 6662877 .5823349 2398228</pre>	.1492527 .5885473 .258483 .1535997	-0.28 -1.13 2.25 -1.56	0.782 0.258 0.024 0.118	2867246 -1.634362 .1571682 4924718	.2042731 .3017865 1.007502 .0128263		
Defenses_Num Vics_AnyHostage D_RaceB cons	<pre>746949975419354365771.15871</pre>	.3710391 .5890277 .7811117 .7728872	-2.01 1.28 1.84 -1.50	0.044 0.200 0.066 0.134	-1.357255 2146709 .1517627 -2.429996	1366449 1.723058 2.721392 .1125764		

To summarize, relevant cases should not be removed from the analysis. However, even when three or four relevant trial reports are excluded, the findings continue to indicate that Black defendants are at least four times as likely to receive the death penalty as non-Black defendants. The exact p-values also remain below the .10 threshold. Table 5 below summarizes these results.

Table 5. Summary: Results of Original Model and Two Variants Excluding Relevant Cases							
Cases Included	Odds Ratio	Exact P-value					
All relevant TRs included (N=77)	4.9	0.039**					
TRs 7, 180 and 216 excluded (N=74)	4.2	0.066*					
TR 7, 180, 216 and 34A excluded (N=73)	4.0	.075*					
* significant at α = .10 ** significant at α = .05	*** significant	at α = .01					

Note: Data entry errors have been corrected and the appropriate logarithmic transformations have been conducted in all of these models.

Interrogatory 23

Do you agree with the above general description of MLEs? If not, please indicate what corrections you would make in the description.

Yes, we are comfortable with the general description provided by the Commissioner of Maximum Likelihood Estimates (MLEs). We would, however, add some relevant context. The last sentence of the description provided by the Commissioner on p. 25 reads: "MLE procedures are generally used for larger datasets for the reason that chance variation may account for the distribution of the bulk of the data in a small dataset, in contrast to a large dataset where it is unlikely that the distribution of the bulk of the bulk of the data is due to chance variation."

To clarify, all data sets, large or small, contain some systematic variation that can be explained and chance or random variation, which cannot. Random variation is different than sampling bias. If researchers draw a sample from the relevant population, they must address the possibility that the distribution of the data in the sample is not fully representative of the population they seek to understand. In this study, however, we are not analyzing a sample of Washington State capital cases. Instead, the data include *all* Washington State aggravated murder proceedings that took place from 1981 to May of 2014 for which trial reports are available. The data thus consist of the population of relevant proceedings rather than a sample of that population; sampling bias cannot account for the distribution of our data.

We used MLE procedures because MLE is an estimation method used for analyzing binary outcomes (e.g., death imposed vs. death not imposed.) In addition, the MLE equation is derived from the probability distribution of the dependent variable, producing a set of parameters for which the probability of the observed data is greatest. The benefit of this technique is that it produces estimates that tend toward the true values and uses the data most efficiently to do so. In statistical terms, MLE generates estimates that are consistent, asymptotically normal, and

asymptotically efficient.¹² These advantages are amplified when data sets are large, but also exist for smaller datasets.

One of the main concerns regarding analysis of small datasets is they may not have sufficient power to identify significant effects where they exist. Clearly, this is not a problem in this analysis. We consistently find, across many model variations, that Black defendants are four times as likely to be sentenced to death in Washington State as similarly situated non-Black defendants.

It is also important to ensure that the findings derived from smaller datasets are robust across numerous model specifications. By robust, we mean that the findings are similar across different model specifications. It is for this reason that we ran numerous models to ensure that the finding regarding the impact of defendant race is consistent rather than aberrational (see Appendix E of our *Updated Report*, pp. 78-89 of our *Response to Scurich*, and the findings reported in this document). The fact that the number of sentencing outcomes analyzed is relatively small also means that these results should be interpreted in conjunction with the descriptive findings. In this case, the descriptive results indicate that 38.8 percent of non-Black defendants, but 64.3 percent of Black defendants, were sentenced to death in special sentencing proceedings in which capital punishment could have been imposed by the judge or jury. The descriptive results also provide evidence that this large discrepancy is not a function of differences in case characteristics (see Table B on p. 17 of our *Response to Evaluation*).

Interrogatory 24

Does identifying and removing a data point that is an outlier address the extent to which chance variation accounts for the distribution of the remaining data for the purposes of MLEs? If yes, please explain.

No, outliers do not have a significant impact on the overall distribution of the data, although they may have an impact on the regression results.

To clarify, a data point may be an outlier with respect to a single variable or with respect a constellation of variables. Outliers with respect to a single variable are comparatively easy to detect. For example, by examining descriptive statistics for the full dataset, including minimum and maximum values, we determined that a handful of proceedings are outliers with respect to the number of victims (see our *Updated Report* at p. 19). In order to retain as complete a dataset as possible, we measured the number of victims in terms of three categories: one victim; two-four victims; or five or more victims. We also constructed a binary variable in which

¹² See Jeffrey Woolridge, INTRODUCTORY ECONOMETRICS: A MODERN APPROACH, 3rd Ed. Mason, Ohio: Thomson Higher Education, 2006, p. 586-594.

cases were coded as involving one or more than one victim. The use of these categories eliminated potential outliers with respect to number of victims.

Diagnostic tests are used to identify data points that are outliers with respect to a constellation of variables. As noted in our *Response to Evaluation*, use of these tests, including examination of Standardized Pearson Residuals, Deviance Residual, and leverage plots, indicated that one case in our dataset may have been an outlier (see fn. 91 on p. 51).¹³

However, removing this case (TR 31) from the analysis had little effect on the results. Specifically, after removing this case, the coefficient for Black defendant was 1.56 (meaning that Black defendants are 4.75 times as likely than non-Black defendants to be sentenced to death after controlling for the other variables in the model) with a p-value of 0.042 (see unaltered output below). When this case is included, the coefficient for Black defendant is 1.58 (meaning that Black defendants are 4.8 times as likely than non-Black defendants to be sentenced to be sentenced to death after controlling for the other variables in the model) with a p-value of 0.039 (see Table 1 of this document). All results presented in this and prior reports include this case.

```
OUTPUT AFTER DROPPING TR 31
```

```
logit DP Sentence InPriors Vics 1Total AppliedAggCir Num LnTotMitCircum
> Defenses Num Vics AnyHostage D RaceB , level(90) ;
Iteration 0: log likelihood = -51.727841
Iteration 1: log likelihood = -39.757487
Iteration 2: log likelihood = -39.344986
Iteration 3: log likelihood = -39.34175
Iteration 4: log likelihood = -39.341749

        Number of obs
        =
        76

        LR chi2(7)
        =
        24.77

        Prob > chi2
        =
        0.0008

        Pseudo R2
        =
        0.2394

Logistic regression
Log likelihood = -39.341749
                                                                                       Pseudo R2
                                                                                                                            0.2394
                                 _____
         DP Sentence | Coef. Std. Err. z P>|z| [90% Conf. Interval]
    ______

      lnPriors
      -.0377231
      .1499017
      -0.25
      0.801
      -.2842895
      .2088433

      Vics_1Total
      -.6689286
      .5932979
      -1.13
      0.260
      -1.644817
      .3069596

      AppliedAggCir_Num
      .5891134
      .2602286
      2.26
      0.024
      .1610754
      1.017151

      LnTotMitCircum
      -.2634829
      .1508145
      -1.75
      0.081
      -.5115507
      -.015415

        Defenses_Num |
        -.7721959
        .3723614
        -2.07
        0.038
        -1.384676

        Vics_AnyHostage |
        .7485637
        .5888958
        1.27
        0.204
        -.2200837

                                                                                                                                   -.159716
                                                                              1.27 0.204 -.2200837 1.717211
               D_RaceB | 1.557717 .7665897
                                                                              2.03 0.042
                                                                                                              .2967894 2.818645
                  _cons | -1.152822 .7754773 -1.49 0.137 -2.428369 .1227246
```

¹³ Out of an abundance of caution, we also assessed the impact of a small number of other data points that showed potential leverage or influence.

Interrogatory 25

Does testing for robustness address the degree to which chance variation accounts for the distribution of the data for purposes of MLEs? If yes, explain.

Robustness and chance variation are separate issues. The robustness of the results is a concern in all regression-based studies, but, as noted above, is of particular concern where datasets are small. As has been made evident, we conducted numerous analyses to ensure that the findings presented in our work are robust across many different model specifications.

Interrogatory 26

Please provide a full description of the methods and associated testing used in selecting which case characteristic variables to include or exclude in analyses of sentencing decisions and the results of such selection methods and testing. The description of methods and associated testing should include, but not be limited to, the variable Judge_ProlongedSuffInd.

In modeling both prosecutorial decisions to file a death notice and jury/judicial decisions to impose a death sentence, we adopted the following strategy: 1) test case characteristics, all but one of which are expected to impact case outcomes; 2) test social characteristics of the defendant found to be relevant in other death penalty studies; 3) test social characteristics of the victim(s) found to be relevant in other death penalty studies; and 4) test other social factors found to be relevant in other studies and that were consistently recorded in the trial reports. Our decisions about variable selection were also influenced by technical issues, including the limit to the number of variables that could be included in a given model at once, the results of diagnostic tests, and the results of sensitivity analyses.

When introducing new variables to the model, we employed a "stepwise" approach, adding one new variable at a time to a base model to examine differences in the results. We first included selected case characteristics to determine their impact, then examined the correlation matrix of these variables to check for potential collinearity. Next, we examined the distributions of each variable and assessed whether variable transformations resulted in improved fit of the data.

Below, we provide a narrative description of the methods and associated testing used in selecting which case characteristics to include in the analyses, and how to measure them, followed by the unaltered statistical output for these models. Because we understand this interrogatory to pertain to the process of selecting variables for the logistic regression models regarding sentencing decisions (as opposed to prosecutorial filing decisions), we focus on the former here. A guide for interpreting STATA output is available in our *Response to Evaluation* at 54-58.

The Model Building Process

After identifying eight main variables that captured potentially relevant case characteristics that would have been known by judges and jurors and for which sufficient data existed,¹⁴ we began by testing a subset of these, consistent with the stepwise approach to model-building. These three variables include the number of aggravating circumstances found, the number of prior convictions, and the total number of mitigating circumstances. Two cases with missing information on prior convictions were dropped from this preliminary analysis, resulting in 80 observations. The output associated with this model appears below. Significant findings are highlighted.

```
BASE MODEL OF CASE CHARACTERISTICS
logit DP_Sentence Priors AppliedAggCir_Num TotMitCircum1, level(90) ;
Iteration 0: log likelihood = -54.548369
Iteration 1: log likelihood = -49.078276
Iteration 2: log likelihood = -49.008453
Iteration 3: log likelihood = -49.008331
Iteration 4: log likelihood = -49.008331
                                                                          Number of obs =
Logistic regression
                                                                                                                80
                                                                          LR chi2(3) = 11.08
Prob > chi2 = 0.0113
                                                                          Prob > chi2
                                                                          Pseudo R2
Log likelihood = -49.008331
                                                                                                 =
                                                                                                          0.1016
       _____
        DP Sentence | Coef. Std. Err. z P>|z| [90% Conf. Interval]

      Priors
      .0893749
      .0605603
      1.48
      0.140
      -.0102379
      .1889876

      AppliedAggCir_Num
      .3892101
      .1828195
      2.13
      0.033
      .0884989
      .6899214

      TotMitCircum1
      -.1152768
      .1244844
      -0.93
      0.354
      -.3200355
      .0894818

      _cons
      -1.287502
      .5817227
      -2.21
      0.027
      -2.244351
      -.3306532
```

In this model, only the number of aggravating circumstances is significantly associated with death sentences (p = .033) when these variables are included in their untransformed form.

Next, we examined each of these variable's distribution in order to determine whether it was appropriate to transform them. Decisions regarding variable transformation are based on a number of factors, including the degree of skew, the range of values involved, issues of collinearity, and the degree to which transforming the variable improves model fit. Histograms for the three independent variables included in this model appear below.

¹⁴ These include: the number of aggravating circumstances, mitigating circumstances, prior convictions, defenses, and victims; the nature of the defendant's plea; and two measures of victim suffering (victim held hostage and prolonged suffering).



Histogram of Number of Prior Convictions: Skewed to the Right, Range = 0 - 25

Histogram of Number of Aggravating Circumstances: Skewed to the Right, Range = 1-12





Histogram of Number of Mitigating Circumstances: Skew to the Right, Range = 0 - 11

All three of these variables showed some signs of skew; we therefore tested a natural log transformation of these variables. In cases in which the associated defendant had zero prior convictions or mitigating circumstances, we added 0.001 to these zeroes, as is standard practice, and then calculated the natural log of the relevant variable.

We also assessed the collinearity among the transformed variables. A correlation matrix of the transformed variables appears below.

CORRELATION MATRIX

```
pwcorr DP Sentence lnPriors LnAppliedAgg LnTotMitCircum, sig
            | DP Sen~e lnPriors LnAppl~g LnTotM~m
DP Sentence |
                1.0000
              0.1034
                        1.0000
   lnPriors |
                0.3612
                0.2825 0.0967
LnAppliedAgg |
                                 1.0000
                0.0101 0.1034
LnTotMitCi~m | -0.2228 -0.0952 -0.0026
                                          1.0000
               0.0442
                         0.3980
                                  0.9815
```

Because the correlation matrix shows no significant correlation between these independent variables, we determined that all could be included in the same model. We tested each of the transformed variables in the model, including one at a time to see if inclusion of the

transformed version of the variable notably changed the results or improved model fit. The output associated with each of these three models appears below.

```
TESTING LOGGED PRIOR CONVICTIONS
logit DP Sentence lnPriors AppliedAggCir Num TotMitCircum1, level(90) ;
Iteration 0: log likelihood = -54.548369
Iteration 1: log likelihood = -50.131102
Iteration 2: log likelihood = -50.097959
Iteration 3: log likelihood = -50.097926
Iteration 4: log likelihood = -50.097926
                                                                    Number of obs =
Logistic regression
                                                                                                      80
                                                                                                8.90
                                                                   LR chi2(3) =
                                                                  Prob > chi2
                                                                                               0.0306
                                                                                        =
Log likelihood = -50.097926
                                                                  Pseudo R2
                                                                                        =
                                                                                                0.0816
             _____
       DP Sentence | Coef. Std. Err. z P>|z|
                                                                                   [90% Conf. Interval]

      InPriors
      .0527723
      .1232335
      0.43
      0.668
      -.1499287
      .2554734

      AppliedAggCir_Num
      .4057318
      .1881323
      2.16
      0.031
      .0962817
      .7151819

      TotMitCircum1
      -.1200149
      .1222348
      -0.98
      0.326
      -.3210732
      .0810434

      __cons
      -.9532644
      .5259274
      -1.81
      0.070
      -1.818338
      -.0881907
```

The results indicate that number of prior convictions is not a significant predictor of death sentences in either its transformed or untransformed state. Because this variable's distribution does indicate significant skew, and involves relatively expansive range of values, we subsequently included the logged version of this variable.

```
TESTING LOGGED AGGRAVATING CIRCUMSTANCES
logit DP Sentence Priors LnAppliedAgg TotMitCircum1, level(90) ;
Iteration 0: log likelihood = -54.548369
Iteration 1: log likelihood = -49.724273
Iteration 2: log likelihood = -49.711064
Iteration 3: log likelihood = -49.711062
                                           Number of obs = 80
LR chi2(3) = 9.67
Prob > chi2 = 0.0215
Logistic regression
Log likelihood = -49.711062
                                            Pseudo R2
                                                          =
                                                               0.0887
                        ------
 DP Sentence | Coef. Std. Err. z P>|z| [90% Conf. Interval]
    ------
     Priors |.0873797.06090571.430.151-.0128013.1875607LiedAgg |.9048161.40921112.210.027.23172371.577909
LnAppliedAgg |
TotMitCircum1 | -.1220763 .1238665 -0.99 0.324 -.3258186
                                                               .081666
  cons | -.9676286 .5156375 -1.88 0.061 -1.815777 -.1194803
```

The number of aggravating circumstances is statistically significant when included as a logged variable and when untransformed. Since the distribution of this variable is only moderately

skewed, we decided to include the untransformed version because it allows for a more straightforward interpretation.

Mitigating circumstances showed signs of significant skew, and logging improves the model fit (i.e., with the correction to TR 25, the Pseudo R² increases from .1016 to .1290). We therefore used the logged version of this variable. Please note that the output shown above includes all data entry error corrections.

Our base model thus includes the logged number of prior convictions and mitigating circumstances as well as the (untransformed) number of aggravating circumstances.

```
BASE MODEL WITH CASE CHARACTERISTICS (MODEL A)
logit DP_Sentence lnPriors AppliedAggCir_Num LnTotMitCircum, level(90) ;
Iteration 0: log likelihood = -54.548369
Iteration 1: log likelihood = -48.797886
Iteration 2: log likelihood = -48.691842
Iteration 3: log likelihood = -48.691734
Iteration 4: log likelihood = -48.691734
                                                                              Number of obs =
Logistic regression
                                                                                                                     80

      LR chi2(3)
      =
      80

      Prob > chi2
      =
      0.0084

Log likelihood = -48.691734
                                                                              Pseudo R2
                                                                                                      =
                                                                                                               0.1074
     _____
                                                                                 P>|z| [90% Conf. Interval]
        DP Sentence | Coef. Std. Err. z
     _____+

      lnPriors
      .047665
      .1249077
      0.38
      0.703
      -.1577899
      .2531199

      pliedAggCir_Num
      .4474499
      .2048967
      2.18
      0.029
      .1104248
      .7844749

      InTotMitCircum
      -.2171533
      .113546
      -1.91
      0.056
      -.4039198
      -.0303868

      _cons
      -1.35109
      .4924975
      -2.74
      0.006
      -2.161176
      -.5410038

AppliedAggCir_Num | .4474499
```

We next assessed the possibility that while the total number of prior convictions is not a significant predictor of sentencing outcomes, the number of specific types of prior convictions might be. The trial reports contain information regarding prior convictions for homicide as well as violent and sex offenses. These variables are also skewed. We tested transformations of these variables. The correlation matrixes for types of prior convictions before and after logarithmic transformation appear below.

CORRELATION MATRICES						
pwcorr DP_Sen	tence Priors	B D_Prior	sHomicide	D_Priors	Violent D_PriorsSex , sig	
	DP_Sen~e	Priors	D_Prio~e	D_Prio~t 1	D_Prio~x	
DP_Sentence	1.0000 					
Priors	 0.1998 0.0755	1.0000				
D_PriorsHo~e	0.1427 0.2065	0.2325 0.0001	1.0000			
D_PriorsVi~t	0.2243 0.0455	0.5625 0.0000	0.5278 0.0000	1.0000		
D_PriorsSex	 0.1560 0.1670 	0.1171 0.0486	-0.0132 0.8248	0.2278 0.0001	1.0000	
. pwcorr DP_S	entence lnPr	riors lnH	omocidePr	iors lnVid	olentPriors lnSexPriors , sig	
	DP_Sen~e 1 +	lnPriors	lnHomo~s	lnViol~s .	lnSexP~s	
DP_Sentence	1.0000 					
lnPriors	0.1034 0.3612	1.0000				
lnHomocide~s	0.1612 0.1531	0.1514 0.0106	1.0000			
lnViolentP~s	0.0629 0.5795	0.5431 0.0000	0.2249 0.0001	1.0000		
lnSexPriors	0.1094 0.3341	0.2022	-0.0021 0.9713	0.2607 0.0000	1.0000	

As is shown above, the number of prior convictions is significantly correlated with the number of prior homicide, violent, and sex offense convictions at high levels. Moreover, the strength of correlations is relatively high. This collinearity exists regardless of whether these variables are transformed. This indicates that only one of these variables should be included in the model at a time.

We assessed whether the number of prior violent convictions is a significant predictor of sentencing outcomes. For reference, we first provide the output obtained when total prior convictions is included, then show the results obtained when the number of prior violent, homicide and sex convictions is included.

```
MODEL A USING LOGGED NUMBER OF (ALL) PRIOR CONVICTIONS
logit DP Sentence lnPriors AppliedAggCir Num LnTotMitCircum, level(90) ;
Iteration 0: log likelihood = -54.548369
Iteration 1: log likelihood = -48.797886
Iteration 2: log likelihood = -48.691842
Iteration 3: log likelihood = -48.691734
Iteration 4: log likelihood = -48.691734
                                         Number of obs = 80
= 11.71
Logistic regression
                                         LR chi2(3) = 11.71
Prob > chi2 = 0.0084
Pseudo R2 = 0.1074
                                         Prob > chi2
Pseudo R2
Log likelihood = -48.691734
     _____
    DP Sentence | Coef. Std. Err. z P>|z| [90% Conf. Interval]
 _____
InPriors.047665.12490770.380.703-.1577899.2531199AppliedAggCir_Num.4474499.20489672.180.029.1104248.7844749
  LnTotMitCircum | -.2171533 .113546 -1.91 0.056 -.4039198 -.0303868
       cons | -1.35109 .4924975 -2.74 0.006 -2.161176 -.5410038
```

MODEL A USING LOGG	ED NUMBER OF V	VIOLENT PRIOR	R CONVIC	TIONS		
logit DP_Sentence	lnViolentPric	ors AppliedAq	ggCir_Nu	m LnTotMi	tCircum, l	evel(90) ;
Iteration 0: log Iteration 1: log Iteration 2: log Iteration 3: log Iteration 4: log	<pre>likelihood = likelihood = likelihood = likelihood = likelihood =</pre>	-54.548369 -48.842185 -48.73727 -48.73718 -48.73718				
Logistic regressio	n		N1 L1 P1	umber of R chi2(3) rob > chi	obs = = 2 =	80 11.62 0.0088
Log likelihood =	-48.73718		P	seudo R2	=	0.1065
DP_Sentence	Coef.	Std. Err.	z	P> z	[90% Co	nf. Interval]
lnViolentPriors AppliedAggCir_Num LnTotMitCircum	<pre>.0224711 .4549994</pre>	.0936624 .2034407 .1141925 5339382	0.24 2.24 -1.90	0.810 0.025 0.057	131589 .120369 40476	9 .1765321 2 .7896296 370291072 1 - 4200508

InHomocidePri							
TUTOROCTOGLIJ	lors Applied	AggCir_N	lum LnTotM	itCircum,	, leve	≥l(90) ;	
likelihood = likelihood = likelihood = likelihood =	-54.548369 -48.287297 -48.165373 -48.16411						
likelinood =	-48.164109	N	Number of R chi2(3)	obs = =	1	80 12.77	
Log likelihood = -48.164109			Prob > chi2 = 0.0052 Pseudo R2 = 0.1170				
Coef.	Std. Err.	Z	P> z	[90% (Conf.	Interval]	
.2406648 .4577821 186735	.2409062 .2012391 .1181576	1.00 2.27 -1.58	0.318 0.023 0.114	15559 .1267 38108	906 733 869	.6369202 .7887909 .0076169	
	<pre>likelihood = likelihood = likelihood = likelihood = likelihood = likelihood = Coef2406648 .45778211867352965373</pre>	<pre>likelihood = -54.548369 likelihood = -48.287297 likelihood = -48.165373 likelihood = -48.16411 likelihood = -48.164109 8.164109 Coef. Std. Err2406648 .2409062 .4577821 .2012391186735 .11815762965373 1.167782</pre>	likelihood = -54.548369 likelihood = -48.287297 likelihood = -48.165373 likelihood = -48.16411 likelihood = -48.164109 8.164109 Coef. Std. Err. z .2406648 .2409062 1.00 .4577821 .2012391 2.27 186735 .1181576 -1.58 2965373 1.167782 -0.25	<pre>likelihood = -54.548369 likelihood = -48.287297 likelihood = -48.165373 likelihood = -48.16411 likelihood = -48.164109 Number of of LR chi2(3) Prob > chi 8.164109 Coef. Std. Err. z P> z .2406648 .2409062 1.00 0.318 .4577821 .2012391 2.27 0.023 186735 .1181576 -1.58 0.114 2965373 1.167782 -0.25 0.800</pre>	<pre>likelihood = -54.548369 likelihood = -48.287297 likelihood = -48.165373 likelihood = -48.164109 Number of obs = LR chi2(3) = Prob > chi2 = 8.164109 Coef. Std. Err. z P> z [90% control = .2406648 .2409062 1.00 0.3181555 .4577821 .2012391 2.27 0.023 .1267 186735 .1181576 -1.58 0.11438100 2965373 1.167782 -0.25 0.800 -2.2173</pre>	<pre>likelihood = -54.548369 likelihood = -48.287297 likelihood = -48.165373 likelihood = -48.16411 likelihood = -48.164109 Number of obs = LR chi2(3) = 1 Prob > chi2 = 0. 8.164109 Pseudo R2 = 0. Coef. Std. Err. z P> z [90% Conf. .2406648 .2409062 1.00 0.3181555906 .4577821 .2012391 2.27 0.023 .1267733 186735 .1181576 -1.58 0.1143810869 2965373 1.167782 -0.25 0.800 -2.217368</pre>	

```
MODEL A USING LOGGED NUMBER OF SEX PRIOR CONVICTIONS
logit DP Sentence lnSexPriors AppliedAggCir Num LnTotMitCircum, level(90) ;
Iteration 0: log likelihood = -54.548369
Iteration 1: log likelihood = -48.857738
Iteration 2: log likelihood = -48.757666
Iteration 3: log likelihood = -48.75757
Iteration 4: log likelihood = -48.75757
                                         Number of obs =
Logistic regression
                                                              80
                                                           11.58
                                                         11.2
0.0090
1062
                                         LR chi2(3) =
                                         Prob > chi2
                                                      =
Log likelihood = -48.75757
                                         Pseudo R2
                                                      =
                                                           0.1062
         _____
    DP Sentence | Coef. Std. Err. z P>|z| [90% Conf. Interval]
      lnSexPriors | .0196884 .1521259 0.13 0.897 -.2305366 .2699133
AppliedAggCir_Num | .4540187 .2076052 2.19 0.029 .1125386 .7954989
  LnTotMitCircum | -.2187364 .1144848 -1.91 0.056 -.4070471 -.0304257
          cons | -1.258491 .8517008 -1.48 0.140 -2.659414 .1424323
```

Because the logged number of violent, homicide, or sex prior convictions are not significant predictors of sentencing outcomes, we determined that controlling for criminal history using the total number of prior convictions is most appropriate as it provides the most comprehensive measure of prior criminal involvement.

We then began to test other case characteristics that would have been known to juries and/or judges. These include: the nature of the defendant's plea (guilty vs. not guilty), the number of defenses offered, and the number of victims.¹⁵ We examined the correlation matrices of these case characteristics as a set (see below).

¹⁵ There are no cases in which there is missing information on the number of victims, number of aggravating circumstances found, or the defendant's plea. Two cases do not have information on the number of prior convictions.

```
CORRELATION MATRIX
pwcorr DP Sentence lnPriors AppliedAggCir Num LnTotMitCircum Defenses Num Vics Num,
sig
            | DP_Sen~e lnPriors Applie~m LnTotM~m Defens~m Vics_Num
DP Sentence |
                1.0000
   lnPriors | 0.1034
                       1.0000
                0.3612
AppliedAgg~m | 0.2906 0.1130
                                1.0000
                0.0081 0.0567
LnTotMitCi~m | -0.2228 -0.0952 -0.0750
                                         1.0000
                       0.3980
                               0.5001
               0.0442
                                                 1.0000
Defenses Num |
               -0.2284 -0.1112 0.0327
                                         0.1879
               0.0403 0.0671 0.5826
                                         0.0909
   Vics Num |
                0.1002 -0.0318
                                 0.0196
                                         0.0650 -0.0334
                                                          1.0000
                0.3702
                       0.5932
                                0.7361
                                         0.5596
                                                  0.5752
```

The correlation matrix indicates very modest correlation between number of defenses and mitigating circumstances (correlation coefficient = .1879) and number of defenses and logged number of priors (correlation coefficient= -.1112). This level of correlation does not pose a problem for including all three variables in the same model.

We also tested whether the number of victims impacts sentencing decisions.



Histogram of Number of Victims, Skew to the Right, Range 1-13

In addition to being skewed to the right, the minimum number of victims is 1 with a maximum of 13 and a mean of 1.94. This indicates that there are a small number of outliers with values that are extreme compared to the other observations. Over one-half (61%) of proceedings involve only one victim and the vast majority (86.6%) include one or two victims.

					Cumulative
		Frequency	Percent	Valid Percent	Percent
Valid	1	50	61.0	61.0	61.0
	2	21	25.6	25.6	86.6
	3	7	8.5	8.5	95.1
	4	2	2.4	2.4	97.6
	13	2	2.4	2.4	100.0
	Total	82	100.0	100.0	

Vics_Num Number of victims

Based on this information, and to avoid unnecessarily dropping cases, we transformed the number of victims into an ordinal variable with three categories: 1) Vics_1Total, 2) Vics_2-4Total, 3) Vics_5PlusTotal. Because just over one-half of the cases involve only one victim, we also created a binary variable: 1) Vics_1Total and 2) Morethan1VicTotal.

Regression results testing these two alternative measures appear below.

MODEL B INLCUDING N VICTIMS	NUMBER OF VIC	TIMS AS THREE	E CATEGOI	RIES: 1 VI	ICTIM, 2-4 VI	CTIMS, 5+
logit DP_Sentence Vics_2_4Total, leve	<pre>lnPriors Appl el(90) ;</pre>	LiedAggCir_Nu	ım LnTotl	MitCircum	Vics_1Total	
Iteration 0: log Iteration 1: log Iteration 2: log Iteration 3: log Iteration 4: log	likelihood = likelihood = likelihood = likelihood =	-54.548369 -48.173443 -48.055823 -48.055641 -48.055641				
Logistic regression Log likelihood = -4	1 18.055641		Nu LI P: P:	umber of c R chi2(5) rob > chi2 seudo R2	bbs = = 2 = 0 = 0	80 12.99 .0235 .1190
DP_Sentence	Coef.	Std. Err.	 Z	P> z	[90% Conf.	Interval]
InPriors AppliedAggCir_Num LnTotMitCircum Vics_1Total Vics_2_4Total	.0811414 .413191 2442835 -1.048636 - 5037135	.1318695 .2102976 .1188293 1.514604 1.516274	0.62 1.96 -2.06 -0.69 -0.33	0.538 0.049 0.040 0.489 0.740	1357647 .0672823 4397403 -3.539939 -2.997762	.2980474 .7590997 0488266 1.442666 1.990335
cons	4691061	1.516853	-0.31	0.757	-2.964108	2.025896

MODEL B INLCUDING N	UMBER OF VI	CTIMS AS TWO	CATEGORI	ES: 1 VIC	TIM, MULTIPL	E VICTIMS
logit DP_Sentence	lnPriors App	pliedAggCir_N	Jum LnTotl	MitCircum	Vics_1Total	, level(90) ;
Iteration 0: log Iteration 1: log Iteration 2: log Iteration 3: log Iteration 4: log	likelihood = likelihood = likelihood = likelihood =	= -54.548369 = -48.227205 = -48.110674 = -48.110491 = -48.110491				
Logistic regression			Ni Li P:	umber of R chi2(4) rob > chi	obs = = 2 =	80 12.88 0.0119
Log likelihood = -4	8.110491		P:	seudo R2	=	0.1180
DP_Sentence	Coef.	Std. Err.	Z	P> z	[90% Conf	. Interval]
InPriors AppliedAggCir_Num LnTotMitCircum Vics_1Total	.0714309 . 4123641 244045 5777524	.1275776 .210538 .1188967 .5374276	0.56 1.96 -2.05 -1.08	0.576 0.050 0.040 0.282	1384155 .0660599 4396128 -1.461742	.2812774 .7586682 0484773 .3062374
_cons	9296028	.6209588	-1.50	0.134	-1.950989	.0917835

As this output shows, the number of victims (measured two different ways) is not a significant predictor of death sentences. For ease of interpretation, we include the number of victims as the binary variable in the model.

We then tested in a step-wise fashion the impact of other case characteristics, including the number of defenses and whether the defendant pled guilty, on sentencing outcomes.



Histogram of Number of Defenses, Mild Skew to the Right, Range = 0-4

					Cumulative
		Frequency	Percent	Valid Percent	Percent
Valid	0	37	45.1	45.7	45.7
	1	25	30.5	30.9	76.5
	2	15	18.3	18.5	95.1
	3	3	3.7	3.7	98.8
	4	1	1.2	1.2	100.0
	Total	81	98.8	100.0	
Missing	System	1	1.2		
Total		82	100.0		

Defenses_Num Number of defenses

The variable number of defenses shows modest skew to the right, but has a limited range (minimum of 0, max of 4) with 98.8% of the observations between 0 and 3. This indicates that variable transformation is not necessary.

Information on number of defenses is missing for one case, reducing the total number of cases analyzed to 79.

```
MODEL C ADDING NUMBER OF DEFENSES
logit DP Sentence InPriors AppliedAggCir Num LnTotMitCircum Vics 1Total
> Defenses_Num, level(90) ;
Iteration 0: log likelihood = -53.684127
Iteration 1: log likelihood = -45.100595
Iteration 2: log likelihood = -44.930182
Iteration 3: log likelihood = -44.929614
Iteration 4: log likelihood = -44.929614
Logistic regression
                                                                                   Number of obs =
                                                                                                                        79
                                                                                                                      17.51
                                                                                  LR chi2(5) =
                                                                                                           = 0.0036
                                                                                 Prob > chi2
Log likelihood = -44.929614
                                                                                  Pseudo R2
                                                                                                                      0.1631
                                                                                                            =
              _____
        DP_Sentence | Coef. Std. Err. z P>|z| [90% Conf. Interval]
   ______

      lnPriors
      .0247077
      .1297988
      0.19
      0.849
      -.1887923
      .2382076

      pliedAggCir_Num
      .5001099
      .2298201
      2.18
      0.030
      .1220895
      .8781303

      LnTotMitCircum
      -.2387313
      .1256659
      -1.90
      0.057
      -.4454332
      -.0320293

      Vics_1Total
      -.6274877
      .5574814
      -1.13
      0.260
      -1.544463
      .2894875

      Defenses_Num
      -.6099021
      .3238293
      -1.88
      0.060
      -1.142554
      -.0772502

      __cons
      -.627894
      .6594722
      -0.95
      0.341
      -1.712629
      .4568411

AppliedAggCir Num |
```

And here is the output after information regarding the nature of the plea is added to the model.

```
MODEL D ADDING NATURE OF PLEA
logit DP Sentence InPriors AppliedAggCir Num LnTotMitCircum Vics 1Total Defenses Num
Plea Guilty, level(90) ;
Iteration 0: log likelihood = -53.684127
Iteration 1: log likelihood = -44.461162
Iteration 2: log likelihood = -44.280709
Iteration 3: log likelihood = -44.279615
Iteration 4: log likelihood = -44.279615
                                          LR chi2(6) = 79

Prob > chi2 = 0.0045

Pseudo R2 = 0.0045
                                          Number of obs =
Logistic regression
Log likelihood = -44.279615
     _____
    DP Sentence | Coef. Std. Err.
                                       z P>|z|
                                                     [90% Conf. Interval]
   _____
lnPriors | .0298853 .1290122 0.23 0.817 -.1823209 .2420915
AppliedAggCir_Num | .4795732 .2294081 2.09 0.037 .1022304 .856916
  LnTotMitCircum | -.2411986 .1261485 -1.91 0.056 -.4486945 -.0337027
    Vics_1Total | -.6413235 .5603693 -1.14 0.252 -1.563049 .280402
    Defenses_Num | -.7723393 .3619775 -2.13 0.033 -1.367739 -.1769393
                                                    -2.007057 .3774039
    Plea_Guilty | -.8148265 .7248246
                                      -1.12 0.261
          cons | -.3048369 .7205419
                                                                .8803491
                                      -0.42
                                             0.672
                                                    -1.490023
```

As this output shows, the number of defenses is a significant predictor of a death sentence: each additional defense is associated with a decrease in the likelihood of receiving a death sentence. However, the nature of defendants' pleas does not appear to be significantly correlated with death sentences when controlling for other variables in the model.

Including these latter two variables does not dramatically impact other relationships in the model: the magnitude, direction, and significance of aggravating circumstances and number of defenses remain consistent across models. The (logged) number of mitigating circumstances consistently has a negative impact on sentencing outcomes; this effect is significant in most models. The consistency of these findings across a variety of models strengthen our confidence in these findings.

Two Measures of Victim Suffering: Victim Held Hostage and Prolonged Suffering

It is conceivable that judges and juries are more likely to impose death sentences when the crime is especially heinous. The model described above includes two measures of heinousness: the number of victims and aggravating circumstances. There were two other potential indicators of heinousness, and, specifically, victim suffering, that might have been included in the analyses. First, the trial reports prompt judges to indicate whether a victim was held

hostage, which undoubtedly compounds the suffering of the victim.¹⁶ The trial reports list two boxes labeled "Yes" or "No" for the judges to check.

					Cumulative
		Frequency	Percent	Valid Percent	Percent
Valid	0 no	54	65.9	66.7	66.7
	1 yes	27	32.9	33.3	100.0
	Total	81	98.8	100.0	
Missing	999	1	1.2		
Total		82	100.0		

Vics_AnyHostage Any victim held hostage

A second option was to create a measure of suffering from Part (4)(h) of the trial report. Part (4)(h) states: "Please describe the nature and extent of any physical harm or torture inflicted upon the victim prior to death." This question is not phrased as a yes or no question, but rather as an open-ended, compound question in which judges were invited to describe any physical harm *or* torture that occurred.

In an effort to create an additional measure of victim suffering, and in the absence of "yes" and "no" boxes, the protocol instructed coders as follows (from the Codebook at 71):

116: *Victim(s): Prolonged suffering*: does the description of the murder indicate that one or more of the victim's suffering was prolonged or allowed to endure over time?

- a. Enter 0 for no.
- b. Enter 1 for yes.

Per these instructions, coders entered a 0 when judges provided no description of physical harm or torture or when the description provided did not indicate that the victim's suffering was prolonged; coders entered a 1 when judges indicated a victim's suffering was endured over a prolonged period of time.

In retrospect, the coding protocol did not capture the compound nature of the question as it appears on the trial report. In addition, the coding protocol did not provide sufficient guidance to coders for determining what constitutes "prolonged." As a result, we did not see this variable as the most reliable measure of victim suffering. Although we did test Prolonged Suffering in the model including only case characteristics, we chose to include Victim Held Hostage as our measure of victim suffering in the models we presented because we have greater confidence in

¹⁶ Part (4)(g) reads: "Was the victim held hostage during the crime?"

this measure. However, in light of the Commissioner's question about the prolonged suffering variable, we present analyses including each of these two measures of victim suffering below.

Below, the output shows that neither measure of victim suffering has a statistically significant impact on sentencing outcomes in capital cases. The number of aggravating circumstances and defenses, and the (logged) number of mitigating circumstances continue to be significant predictors of sentencing outcomes when Victim Held Hostage is added to the model.

MODEL E ADDING VICTIM HELD HOSTAGE

logit DP Sentence InPriors AppliedAggCir Num LnTotMitCircum Vics 1Total Defenses Num Plea Guilty Vics AnyHostage, level(90) ; Iteration 0: log likelihood = -53.138727 Iteration 1: log likelihood = -42.747574Iteration 2: log likelihood = -42.426599Iteration 3: log likelihood = -42.42515Iteration 4: log likelihood = -42.42515Logistic regression Number of obs = 78

 Number of obs
 =
 /8

 LR chi2(7)
 =
 21.43

 Prob > chi2
 =
 0.0032

 Prob > chi2 Log likelihood = -42.42515Pseudo R2 = 0.2016 DP_Sentence | Coef. Std. Err. z P>|z| [90% Conf. Interval] _____ lnPriors | -.0454251 .137801 -0.33 0.742 -.2720876 .1812375 AppliedAggCir_Num | .5196788 .2425385 2.14 0.032 .1207384 .9186192 LnTotMitCircum | -.2976283 .1392095 -2.14 0.033 -.5266075 -.0686491 Vics 1Total | -.6346856 .5725699 -1.11 0.268 -1.576479 .307108 Defenses Num | -.7950692 .3715923 -2.14 0.032 -1.406284 -.1838542 Plea_Guilty | -.6789026 .7613904 -0.89 0.373 -1.931278 .5734731 Vics AnyHostage | .8627311 .5715877 1.51 0.131 -.077447 1.802909 _cons | -.6152901 .7579476 -0.81 0.417 -1.862003 .6314228

MODEL F SUBSTITUTING	PROLONGED SUFF	ERING FOR V	ICTIM HEI	D HOSTAGE	1		
logit DP_Sentence ln1 Plea_Guilty Judge_Proi	Priors Applied longSuffInd, l	lAggCir_Num .evel(90) ;	LnTotMitC	Circum Vic	s_1Tota	al Def	enses_Num
Iteration 0: log lil Iteration 1: log lil Iteration 2: log lil Iteration 3: log lil Iteration 4: log lil	kelihood = -53 kelihood = -42 kelihood = -42 kelihood = -42 kelihood = -42	.684127 .982838 .746661 .745544 .745544					
Logistic regression Log likelihood = -42.7	745544		Numbe LR ch Prob Pseuc	er of obs hi2(7) > chi2 do R2	= = =	21. 0.00 0.20	79 88 27 38
DP_Sentence	Coef.	Std. Err.	Z	P> z	[90%]	Conf.	Interval]
InPriors AppliedAggCir_Num LnTotMitCircum Vics_1Total Defenses_Num Plea_Guilty Judge_ProlongSuffInd	 0075088 .5015173 2389256 679632 8678216 7875189 1.379501 4210272 	.1324349 .2391731 .1340487 .5728606 .3771392 .736982 .9385685 7252897	-0.06 2.10 -1.78 -1.19 -2.30 -1.07 1.46	0.955 0.036 0.075 0.235 0.021 0.285 0.144	2253 .1081 459 -1.621 -1.48 -1.999 .0378	3448 125 9416 904 3816 9746 3586	.2103272 .8949221 -0184352 .2626398 -2474828 .4247087 2.721143 7784066
_cons	4310373	.1332091	-0.59	0.000	-1.040	7401	. / /04000

As shown above, neither measure of victim suffering is a significant predictor of sentencing outcomes (p-values = .131 and .144). In addition, neither measure has a substantive impact on the coefficients and significance of other independent variables included in the model. The number of aggravating circumstances continues to have a significant, positive association with the imposition of death sentences in either case, and the number of mitigating circumstances and defenses continue to have a significant, negative association with death sentences.

Table 6 below displays a summary of the models discussed above, listing the number of cases included, the model fit metrics, odds ratios, and exact p-values of the variables included.

Table 6. Summary of Models	with Case Chara	cteristics				
	Model A	Model B	Model C	Model D	Model E	Model F
	Base Model					
Ν	80	80	79	79	78	79
Prob > chi2	.0084	.0119	.0036	.0045	.0032	.0027
Pseudo R2	.1074	.1180	.1631	.1752	.2016	.2038
	Odds Ratio	Odds Ratio	Odds Ratio	Odds Ratio	Odds Ratio	Odds Ratio
	(P-Value)	(P-Value)	(P-Value)	(P-Value)	(P-Value)	(P-Value)
Number of Prior Convictions	1.05	1.07	1.03	1.03	.956	.993
(logged)	(0.703)	(0.576)	(0.849)	(0.817)	(0.742)	(0.955)
Number of Aggravating	1.56	1.51	1.65	1.62	1.68	1.65
Circumstances	(0.029)	(0.050)	(0.030)	(0.037)	(0.032)	(0.036)
Number of Mitigating	0.805	.783	.788	.786	.743	.787
Circumstances (logged)	(0.056)	(0.040)	(0.053)	(0.056)	(0.033)	(0.075)
1 Victim		.56	.534	.527	.530	.507
(referent: more than 1 victim)		(0.282)	(0.260)	(0.252)	(0.268)	(0.235)
Number of Defenses			.543	.462	.452	.420
			(0.060)	(0.033)	(0.032)	(0.021)
Pled Guilty (referent: plead				.443	.507	.455
not guilty)				(0.261)	(0.373)	(0.285)
Victim Held Hostage					2.37	
					(0.131)	
Prolonged Suffering Indicated						3.97
by the Judge						(0.144)

Notes: All statistically significant results are bolded. Odds Ratio = 1 indicates no effect; Odds Ratio < 1 indicates a negative effect; Odds Ratio > 1 indicates positive effect.

In this table, all statistically significant results are bolded. The results show that the number of aggravating circumstances, mitigating circumstances, and defenses consistently reach significance at an alpha level of .10. That is, these findings are robust across numerous model specifications.

Given the small number (82, with several dropped due to missing data) of cases in the population under study, it is inappropriate to include more than seven variables in our models at once. In order to test whether race of the defendant or other factors significantly predict death sentences, a model with fewer variables or a more parsimonious version of the base model must be established.

Establishing a Model that Controls for All Relevant Case Characteristics

Given the consistency of the finding that the number of aggravating circumstances, mitigating circumstances, and defenses are significantly correlated with death sentence, these variables should be included in the control model for testing other variables. We next test to see if the relationships among the other four independent variables change when they are substituted out for each other while also including some measure of victim suffering.

First, we test the significance of guilty pleas when including the two different measures of victim suffering. The output appears below.

MODEL G INCLUDING H	PLEA AND VICT	IM HELD HOST	AGE			
logit DP_Sentence A Vics_AnyHostage, le	AppliedAggCir_ evel(90) ;	_Num LnTotMi	tCircum	Defenses_N	Num Plea_Guil	ty
Iteration 0: log Iteration 1: log Iteration 2: log Iteration 3: log Iteration 4: log	<pre>likelihood = likelihood = likelihood = likelihood = likelihood =</pre>	-54.548369 -43.980881 -43.709737 -43.708796 -43.708796				
Logistic regression Log likelihood = -4	1 13.708796		N L F F	Number of d R chi2(5) Prob > chi2 Pseudo R2	bbs = = 2 = 0 = 0	80 21.68 .0006 .1987
DP_Sentence	Coef.	Std. Err.	Z	₽> z	[90% Conf.	Interval]
AppliedAggCir_Num LnTotMitCircum Defenses_Num Plea_Guilty Vics_AnyHostage cons	.5503119 2603829 8090715 7466242 .8782137 -1.054217	.2294586 .129725 .357461 .7410054 .5561479 6359766	2.40 -2.01 -2.26 -1.01 1.58 -1.66	0.016 0.045 0.024 0.314 0.114 0.097	.1728862 4737615 -1.397043 -1.96547 0365682 -2.100306	.9277377 0470042 2211005 .4722213 1.792996 - 0081289

MODEL H SUBSTIT	UTING P	ROLONGED SUFE	TERING FOR VI	ICTIM HEL	D HOSTAGE			
logit DP_Senten Judge_ProlongSu	ce Appl ffInd,	iedAggCir_Num level(90) ;	ı LnTotMitCi	rcum Defe	nses_Num	Plea_G	uilty	
Iteration 0: Iteration 1: Iteration 2: Iteration 3: Iteration 4:	log lik log lik log lik log lik log lik	elihood = -55 elihood = -4 elihood = -44 elihood = -44 elihood = -44	0.097186 4.50221 2.284795 2.284045 2.284045					
Logistic regres Log likelihood	sion = -44.2	84045		Numbe LR ch Prob Pseud	r of obs i2(5) > chi2 o R2	= = =	21.0 0.000 0.190	31 63 06 63
DP_Sen	tence	Coef.	Std. Err.	Z	P> z	[90%]	Conf.	Interval]
AppliedAggCi LnTotMitC Defense Plea_G Judge_ProlongSu	r_Num ircum s_Num uilty ffInd _cons	.5556942 2092867 9209201 8415821 1.309944 8805813	.228948 .1273068 .3700266 .7215402 .8950166 .6000578	2.43 -1.64 -2.49 -1.17 1.55 -1.47	0.015 0.100 0.013 0.243 0.103 0.142	.1793 418 -1.53 -2.03 .003 -1.86	1082 6877 2956 2841 7182 7589	.9322801 .0001144 3122804 .3452458 2.61617 .106426

Below we test number of prior convictions (logged) with two different measures of victim suffering.

```
MODEL I INCLUDING PRIOR CONVICTIONS AND VICTIM HELD HOSTAGE
logit DP Sentence AppliedAggCir Num LnTotMitCircum Defenses Num lnPriors
Vics AnyHostage, level(90) ;
Iteration 0: log likelihood = -53.138727
                     log likelihood = -43.727982
Iteration 1:
Iteration 2: log likelihood = -43.43278
Iteration 3: log likelihood = -43.431995
Iteration 4: log likelihood = -43.431995
                                                                               Number of obs = 78
LR chi2(5) = 19.41
Prob > chi2 = 0.0016
Logistic regression
Log likelihood = -43.431995
                                                                               Pseudo R2
                                                                                                                 0.1827
                                                                                                        =
               _____
       DP Sentence | Coef. Std. Err. z P>|z| [90% Conf. Interval]

      AppliedAggCir_Num |
      .5793639
      .2386559
      2.43
      0.015
      .1868098
      .971918

      InTotMitCircum |
      -.2752854
      .1331734
      -2.07
      0.039
      -.4943362
      -.0562347

      Defenses_Num |
      -.6762022
      .3316308
      -2.04
      0.041
      -1.221686
      -.130718

      InPriors |
      -.0869814
      .1332577
      -0.65
      0.514
      -.3061708
      .132208

      Vics_AnyHostage |
      .7945714
      .5604813
      1.42
      0.156
      -.1273382
      1.716481

      _cons |
      -1.282906
      .597073
      -2.15
      0.032
      -2.265004
      -.3008084
```

MODEL J INCLUDING PRIOR CONVICTIONS AND SUBSTITUTING PROLONGED SUFFERING FOR VICTIM HELD HOSTAGE logit DP_Sentence AppliedAggCir_Num LnTotMitCircum Defenses_Num lnPriors Judge ProlongSuffInd, level(90); Iteration 0: log likelihood = -53.684127 Iteration 1: log likelihood = -44.238541 Iteration 2: log likelihood = -44.031064 Iteration 3: log likelihood = -44.030662 Iteration 4: log likelihood = -44.030662 Number of obs = Logistic regression 79 Number of obs = 79 LR chi2(5) = 19.31 Prob > chi2 = 0.0017 Prob > chi2 = 0.1798 Log likelihood = -44.030662Pseudo R2 _____ DP Sentence | Coef. Std. Err. z P>|z| [90% Conf. Interval] _____ AppliedAggCir Num | .5687897 .2336428 2.43 0.015 .1844814 .953098 LnTotMitCircum | -.2006527 .1262117 -1.59 0.112 -.4082525 .0069472 Defenses_Num | -.723918 .334463 -2.16 0.030 -1.274061 -.1737753 lnPriors | -.0409044 .1306145 -0.31 0.754 -.2557461 .1739373 ngSuffInd | 1.327467 .8949449 1.47 0.123 .0445416 2.610393 Judge_ProlongSuffInd | 1.327467 .8949449 1.47 0.123 _cons | -1.221101 .5503326 -2.22 0.026 -2.126317 -.3158841

Below, we test number of one victim compared to multiple victims with the two different measures of victim suffering.

MODEL K INCLUDING	NUMBER OF VICT	TIMS AND VIC	TIM HELD	HOSTAGE		
logit DP_Sentence ; > Vics_AnyHostage,	AppliedAggCir_ level(90) ;	_Num LnTotMi	tCircum	Defenses_1	Num Vics_2	lTotal
Iteration 0: log Iteration 1: log Iteration 2: log Iteration 3: log Iteration 4: log	likelihood = likelihood = likelihood = likelihood =	-54.548369 -43.68075 -43.337332 -43.336339 -43.336338				
Logistic regression	n 43.336338		N L P P	umber of R chi2(5) rob > chi seudo R2	obs = = 2 = =	80 22.42 0.0004 0.2055
DP_Sentence	Coef.	Std. Err.	 Z	 ₽> z	[90% Cor	nf. Interval]
AppliedAggCir_Num LnTotMitCircum Defenses_Num Vics_1Total Vics_AnyHostage	<pre>.5274189</pre>	.2336575 .1388158 .3306395 .5580146 .5572725 .6974202	2.26 - 2.22 - 2.02 -1.32 1.58	0.024 0.026 0.044 0.186 0.114	.1430864 5367884 -1.211311 -1.655520 0355183	4 .9117513 4080125 11236033 6 .1801787 1 1.797745 7 3342211

MODEL L INCLUDING NUMBER OF VICTIMS AND SUBSTITUTING PROLONGED SUFFERING FOR VICTIM HELD HOSTAGE logit DP_Sentence AppliedAggCir_Num LnTotMitCircum Defenses_Num Vics_1Total Judge ProlongSuffInd, level(90); Iteration 0: log likelihood = -55.097186 Iteration 1: log likelihood = -44.25512Iteration 2: log likelihood = -43.999909 Iteration 3: log likelihood = -43.999472 Iteration 4: log likelihood = -43.999472 Number of obs = Logistic regression 81 Number of obs = 81 LR chi2(5) = 22.20 Prob > chi2 = 0.0005 = 0.2014 Log likelihood = -43.999472Pseudo R2 _____ DP Sentence | Coef. Std. Err. z P>|z| [90% Conf. Interval] ______ AppliedAggCir Num | .5231211 .2331648 2.24 0.025 .1395991 .906643 LnTotMitCircum | -.242838 .1342841 -1.81 0.071 -.4637158 -.0219603 Defenses_Num | -.7417002 .3363259 -2.21 0.027 -1.294907 -.1884934 Vics_1Total | -.7756938 .5576442 -1.39 0.164 -1.692937 .1415492 olongSuffInd | 1.376672 .8535412 1.60 0.112 .0655444 2.687799 Judge_ProlongSuffInd | 1.376672 .8535412 1.60 0.112 _cons | -.6518518 .6637885 -0.98 0.326 .4399831 -1.743687

Table 7 below shows the summary results of the models described above.

Table 7. Establishing Control	Table 7. Establishing Control Variables for Case Characteristics, Testing Measure of Victim Suffering						
	Model G	Model H	Model I	Model J	Model K	Model L	
Ν	80	81	78	79	80	81	
Prob > chi2	.0006	.0006	.0016	.0017	.0004	.0005	
Pseudo R2	.1987	.1963	.1827	.1798	.2055	.2014	
	Odds Ratio	Odds Ratio	Odds Ratio	Odds Ratio	Odds Ratio	Odds Ratio	
	(P-Value)	(P-Value)	(P-Value)	(P-Value)	(P-Value)	(P-Value)	
Number of Aggravating	1.73	1.74	1.78	1.77	1.69	1.69	
Circumstances	(0.016)	(0.015)	(0.015)	(0.015)	(0.024)	(.025)	
Number of Mitigating	.771	.811	.759	.818	.735	.784	
Circumstances (logged)	(0.045)	(0.100)	(0.039)	(0.112)	(0.026)	(0.071)	
Number of Defenses	.445	.398	.509	.485	.513	.476	
	(0.024)	(0.013)	(0.041)	(0.030)	(0.044)	(0.027)	
Pled Guilty (referent: pled	.474	.431					
not guilty)	(0.314)	(0.243)					
Number of Priors (logged)			.917	.960			
			(0.514)	(0.754)			
One Victim (referent: more					.478	.460	
than 1 victim)					(0.186)	(0.164)	
Victim Held Hostage	2.41		2.21		2.41		
	(0.114)		(0.156)		(.114)		
Prolonged Suffering		3.671		3.77		3.96	
Indicated by the Judge		(0.103)		(0.123)		(0.112)	

Note: All statistically significant results are bolded. Odds Ratio = 1 indicates no effect; Odds Ratio < 1 indicates a negative effect; Odds Ratio > 1 indicates a positive effect.

In summary, after testing a variety of constellations of variables together, the results indicate that neither measure of victim suffering is a significant predictor of death sentences and neither affects the overall pattern of results. Because we are less confident in the variable Prolonged Suffering, we included Victim Held Hostage as the measure of victim suffering in the model controlling for case characteristics. It also worth noting that the number of aggravating circumstances and victims also capture aspects of the heinousness of the crime.

Based on the analyses shown above, we identified the following variables to be included in the model that controls for case characteristics: logged number of prior convictions, whether one victim or multiple victims were involved, the number of aggravating circumstances, the logged number of mitigating circumstances, the number of defenses presented, the nature of the defendant's plea, and whether the victim was held hostage. The results of this model (with data errors corrected and including Trial Report 34A) is presented below in Table 8. The unaltered statistical output follows. This table contains the same variables as Table 6 in the *Updated Report* (at 31).

Special Sentencing Proceedings, December 1981 - May 2014							
N= 78	Deat	h Penalty Impo	sed	Pseudo R ² = 0.2016			
				LR chi2(7) = 21.43			
				Prob > chi2 = 0.0032			
Variable	Coefficient	Exact	Odds	90% Confidence			
		P-Value	Ratio	Interval			
Prior Convictions (In)	-0.045	0.742	0.956	272, .181			
1 Victim	-0.635	0.268	0.530	-1.58, .307			
Pled Guilty	-0.679	0.373	0.507	-1.93, .573			
Aggravating Circumstances	0.520	0.032	1.68**	.121, .919			
Mitigating Circumstances (In)	-0.298	0.033	0.743**	527,069			
Defenses	-0.795	0.032	0.452**	-1.41,184			
Victim Held Hostage	0.863	0.131	077, 1.80				

 Table 8. Model M: Impact of Case Characteristics on Capital Sentencing Outcomes in Cases With

 Special Sentencing Proceedings, December 1981 - May 2014

* significant at α = .10

** significant at α = .05

*** significant at α = .01

MODEL M OUTPUT ASS	OCIATED WITH 1	ABLE 8				
logit DP_Sentence Defenses_Num Vics	lnPriors Vics_ _AnyHostage, l	1Total Plea evel(90) ;	_Guilty A	AppliedAg	gCir_Num LnTo	tMitCircum
Iteration 0: log Iteration 1: log Iteration 2: log Iteration 3: log Iteration 4: log	<pre>likelihood = likelihood = likelihood = likelihood = likelihood =</pre>	-53.138727 -42.747574 -42.426599 -42.42515 -42.42515				
Logistic regressio	n		Nu LF Pi	umber of R chi2(7) cob > chi	obs = = 2 = 0	78 21.43 .0032
Log likelihood =	-42.42515		Ps	seudo R2	= 0	.2016
DP_Sentence	Coef.	Std. Err.	Z	P> z	[90% Conf.	Interval]
lnPriors Vics_1Total Plea Guilty	0454251 6346856 6789026	.137801 .5725699 .7613904	-0.33 -1.11 -0.89	0.742 0.268 0.373	2720876 -1.576479 -1.931278	.1812375 .307108 .5734731
AppliedAggCir_Num LnTotMitCircum Defenses_Num	.5196788 2976283 7950692	.2425385 .1392095 .3715923	2.14 -2.14 -2.14	0.032 0.033 0.032	.1207384 5266075 -1.406284	.9186192 0686491 1838542
Vics_AnyHostage cons	.8627311 6152901	.5715877 .7579476	1.51 -0.81	0.131 0.417	077447 -1.862003	1.802909 .6314228

Interrogatory 27

Please provide a full description of the method and any associated testing used in selecting which variable to remove from those included in the model shown in Table 6 in order to add the race of defendant variable to the model as shown in Table 7.

As described in our response to Interrogatory 26, we first identified and tested a model that includes potentially relevant case characteristics that would have been known to the judge or jury. Of these variables, three showed a consistent and significant relationship to decisions to impose death: the number of aggravating circumstances, the number of defenses, and the (logged) number of mitigating circumstances (see Table 8 above). This table contains the same variables as Table 6 in the *Updated Report* at p. 31.

Next we sought to test whether judges and juries are more likely to impose a death sentence when the defendant is Black, controlling for relevant case characteristics. As a reminder, given that the dataset analyzed includes 77 special sentencing proceedings with no missing values, we were limited to seven or fewer explanatory variables in the regression model. We elected to drop defendant plea in order to accommodate defendant race, for several reasons (please see our response to Interrogatory 28 for an explanation of this decision).

Table 9 below shows the results of the model dropping defendant plea and adding defendant race. This table contains the same elements as Table 7 in the *Updated Report* at 32. The unaltered statistical output follows.

Table 9. Impact of Legally Relevant Case Characteristics and Defendant Race on CapitalSentencing Outcomes in Cases with Special Sentencing Proceedings, December 1981 - May2014

N= 77	Deatl	n Penalty Impo	sed	Pseudo R ² = 0.2377
				LR chi2(7) = 24.84
				Prob > chi2 = 0.0008
Variable	Coefficient	Exact	Odds	90% Confidence
		P-Value	Ratio	Interval
Prior Convictions (In)	-0.091	0.510	0.913	319, .137
1 Victim	-0.722	0.221	0.486	-1.69, .248
Aggravating circumstances	0.630	0.016	1.88**	.198, 1.06
Mitigating Circumstances (In)	-0.258	0.089	0. 773*	507,009
Defenses	-0.794	0.034	0. 452**	-1.41,178
Victim Held Hostage	0.717	0.222	2.05	248, 1.68
Black Defendant	1.58 0.039 4.86 **			.319, 2.84

* significant at α = .10

** significant at α = .05

*** significant at α = .01

UNALTERED OUTPUT AS	SSOCIATED WITH	I TABLE 9				
logit DP_Sentence 1	lnPriors Vics_	1 Total A	ppliedAgg	Cir_Num L	nTotMitCircum	n Defenses_Num
Vics AnyHostage D H	RaceB, level(—		—
Iteration 0: log	likelihood =	-52.583924				
Iteration 1: log	likelihood =	-40.664023				
Iteration 2: log	likelihood =	-40.170014				
Iteration 3: log	likelihood =	-40.166274				
Iteration 4: log	likelihood =	-40.166273				
_						
Logistic regression	n		N	umber of	obs =	77
			L	R chi2(7)	=	24.84
			P	rob > chi	2 = 0	.0008
Log likelihood = -4	40.166273		P	seudo R2	= 0	.2361
DP_Sentence	Coef.	Std. Err.	Z	P> z	[90% Conf.	Interval]
	+					
lnPriors	0913328	.1385858	-0.66	0.510	3192863	.1366206
Vics_1Total	7215931	.5896597	-1.22	0.221	-1.691497	.2483108
AppliedAggCir_Num	.6299845	.2624197	2.40	0.016	.1983425	1.061627
LnTotMitCircum	2575945	.1513135	-1.70	0.089	506483	0087059
Defenses_Num	7935932	.3740099	-2.12	0.034	-1.408785	1784016
Vics_AnyHostage	.7169782	.5866349	1.22	0.222	2479504	1.681907
D_RaceB	1.581795	.767834	2.06	0.039	.3188205	2.84477
_cons	-1.114036	.7731921	-1.44	0.150	-2.385824	.1577521

As Table 9 and associated output shows, substituting defendant race for defendant plea improves the fit of the model (Prob > chi2 = .0008 compared to .0032; Pseudo R² =2361 compared to .2016.) In this model, the number of aggravating circumstances, the (logged) number of mitigating circumstances and the number of defenses continue to be statistically significant. Defendant race is also statistically significant (p=.039). When the coefficient for defendant race is expressed as an odds ratio, the results show that Black defendants are 4.9 (4.86) times as likely as non-Black defendants to be sentenced to death, controlling for the other (legally relevant) variables included in the model.

Model N shows the results of the most parsimonious model that includes only the variables that have been shown to be significant predictors of sentencing outcomes: defendant race, number of aggravating circumstances, the (logged) number of mitigating circumstances, and the number of defenses. This is the base model used to test the impact of other social factors presented in the *Response to Evaluation* Tables G and H, at p. 36-38. This is also the base model used to show the impact of defendant race tested across 13 models summarized in *Response to Evaluation* Table I at p. 39-40.

MODEL N PARSIMONIOUS MODEL

logit DP_Senter	nce D_1	RaceB Applie	edAggCir_Num	LnTotMi	tCircum D	efenses_Num ,	level(90) ;
Iteration 0: Iteration 1: Iteration 2: Iteration 3: Iteration 4:	log 1 log 1 log 1 log 1 log 1	ikelihood = ikelihood = ikelihood = ikelihood = ikelihood =	-54.548369 -43.7756 -43.508641 -43.507072 -43.507072				
Logistic regre: Log likelihood	= -43	.507072		Nu LI P: P:	umber of R chi2(4) rob > chi seudo R2	obs = = .2 = 0 = 0	80 22.08 .0002 .2024
DP_Senter	nce	Coef.	Std. Err.	Z	P> z	[90% Conf.	Interval]
D_Rad AppliedAggCir_1 LnTotMitCirc Defenses_1 	ceB Num cum Num ons	1.578166 .6296285 1556172 7689575 -1.389417	.7229784 .235125 .1300711 .3462735 .5818862	2.18 2.68 -1.20 -2.22 -2.39	0.029 0.007 0.232 0.026 0.017	.388972 .2428823 3695651 -1.338527 -2.346535	2.767359 1.016375 .0583307 1993884 4322999

In light of the Commissioner's previous question about the prolonged suffering variable, we reiterate that we have significant concerns about the validity of this measure. Nonetheless, we provide regression results for a model in which this measure of victim suffering is included in place of Victim Held Hostage. Please see Table 10 and associated output below.¹⁷ As these results indicate, the race of the defendant continues to have a large and significant impact on

¹⁷ Associated output for model including Victim Held Hostage appears beneath Table 9.

sentencing outcomes when this measure of victim suffering is included in place of Victim Held Hostage.

Table 10. Testing Black Defendant, Comparing Different Measures of Victim Suffering								
	Including Victim Held	Including Prolonged Suffering						
	Hostage	Indicated by Judge						
Ν	77	78						
Prob > chi2	.0008	.0011						
Pseudo R2	.2361	.2274						
	Odds Ratio	Odds Ratio						
	(P-Value)	(P-Value)						
Black Defendant	4.86**	4.24*						
	(0.039)	(0.064)						
Number of Aggravating	1.88**	1.81**						
circumstances	(0.016)	(0.020)						
Number of Mitigating	.773*	.819						
Circumstances (logged)	(0.089)	(0.160)						
Number of Defenses	.452**	.447**						
	(0.034)	(0.030)						
Number of Priors (logged)	.913	.967						
	(0.510)	(0.800)						
1 Victim	.486	.461						
(referent: multiple	(0.221)	(0.186)						
victims)								
Victim Held Hostage	2.05							
	(0.222)							
Prolonged Suffering		2.57						
Indicated by Judge		(0.264)						
* significant at α = .10	** significant at $\alpha = .05$	*** significant at α = .01 Note:						

All statistically significant results are bolded. Odds Ratio = 1 indicates no effect; Odds Ratio < 1 indicates a negative effect; Odds Ratio > 1 indicates positive effect.

OUTPUT ASSOCIATED WITH TABLE 10 ABOVE, SUBSTITUTING PROLONGED SUFFERING FOR VICTIM HELD HOSTAGE logit DP Sentence D RaceB AppliedAggCir Num LnTotMitCircum Defenses Num lnPriors Vics 1Total Judge ProlongSuffInd, level(90) ; Iteration 0: log likelihood = -53.138727 Iteration 1: log likelihood = -41.411662 Iteration 2: log likelihood = -41.055936 Iteration 3: log likelihood = -41.054392 Iteration 4: log likelihood = -41.054392 Number of obs = Logistic regression 78 Number of obs = 78 LR chi2(7) = 24.17 Prob > chi2 = 0.0011 Prob > chi2 = 0.2274 Log likelihood = -41.054392Pseudo R2 _____ DP Sentence | Coef. Std. Err. z P>|z| [90% Conf. Interval] _____+ D_RaceB | 1.443602 .7785106 1.85 0.064 .1630657 2.724138 AppliedAggCir_Num | .5934399 .2558277 2.32 0.020 .1726407 1.014239 LnTotMitCircum | -.1993755 .1420597 -1.40 0.160 -.433043 .0342919 Defenses_Num | -.8057725 .3703869 -2.18 0.030 -1.415005 -.1965403 InPriors | -.0338136 .1331936 -0.25 0.800 -.2528976 .1852704
 Vics_1Total
 -.7734375
 .5850693
 -1.32
 0.186
 -1.735791
 .1889159

 Judge_ProlongSuffInd
 .9447178
 .8461125
 1.12
 0.264
 -.4470134
 2.336449

 _cons
 -.9264907
 .7294082
 -1.27
 0.204
 -2.12626
 .2732791

Interrogatory 28

Please explain the basis for your statement that the model that omitted this variable [nature of defendant's plea] and added race of defendant is a model that still "included (or controlled for) all relevant case characteristics."

As noted in our response to Interrogatory 27, there are four case characteristics that did not show a significant relationship to sentencing outcomes during model testing. These include: the (logged) number of prior convictions, the number of victims (measured as one vs. more than one victim), victim suffering (measured as whether the victim was held hostage or as prolonged suffering), and whether the defendant entered a guilty plea. We determined that one of these non-significant variables should be removed in order to create space to test the impact of defendant race.

When selecting a single case characteristic to be removed from the model, we chose to remove defendant plea, for several reasons. First, as noted above, defendant plea consistently showed no significant relationship to the decision to impose death (see the analyses presented in response to Interrogatory 26). Second, unlike the other case characteristics, a defendant's plea is not a pre-existing characteristic of either the crime or the defendant. For this reason, it is qualitatively different from the other case characteristics and arguably unrelated to the question of culpability. Finally, the decision to remove this variable was also based on our

understanding that the nature of the defendant's plea ("guilty" or "not guilty") cannot lawfully be the basis of the decision to impose a death sentence, unlike the other case characteristics tested.¹⁸ Therefore, when deciding which category to remove from the model, we removed the category that was less legally relevant than other factors such as number of victims, prior convictions, and victim held hostage, which are proper considerations for imposing death or a life without parole sentence.

It is worth noting, however, that substituting other non-significant case characteristics and including the nature of the plea in their stead does not alter the pattern of results: in all of these models, the same three case characteristics (number of aggravating circumstances, (logged) number of mitigating circumstances, and number of defenses) remain significant, as does race of defendant. The odds ratios for defendant race ranges from 4.2 to 4.9 in these models. Please see the output shown below.

MODIFIED TABLE 7 F	ROM UPDATED RE	EPORT, DROPP	ING PLEA			
logit DP Sentence I) RaceB Applie	edAggCir Num	LnTot.Mit	tCircum D	efenses Num 1	InPriors
Vics 1Total Vics A	vHostage. let	vel(90) :	2011000112	00110um 2		
	19110000ag0, 10	.01()0) /				
Iteration 0. log	likelihood =	-52 583924				
Iteration 1: log	likelihood =	-40 664023				
Iteration 2. log	likelihood -	-40.170014				
Iteration 2. log	likelihood -	-40.170014				
Iteration 5. 10g	likelihood -	40.100274				
iteration 4: log	likelinood -	-40.1002/3				
- ! - ! ! - !			27		- 1	77
Logistic regression	1		NI	umber or	ods =	11
			11	K CN12(/)	=	24.84
			Pi	rob > chi	2 = (0.0008
Log likelihood = -4	40.166273		Ps	seudo R2	= (0.2361
DP_Sentence	Coef.	Std. Err.	Z	P> z	[90% Conf.	. Interval]
D RaceB	1.581795	.767834	2.06	0.039	.3188205	2.84477
AppliedAggCir Num	. 6299845	.2624197	2.40	0.016	.1983425	1.061627
LnTotMitCircum	2575945	.1513135	-1.70	0.089	506483	0087059
Defenses Num	7935932	.3740099	-2.12	0.034	-1.408785	1784016
lnPriors	0913328	.1385858	-0.66	0.510	3192863	.1366206
Vics 1Total	7215931	.5896597	-1.22	0.221	-1.691497	.2483108
Vics AnyHostage	.7169782	.5866349	1.22	0.222	2479504	1.681907
cons	-1 114036	7731921	-1 44	0 150	-2 385824	1577521
	1 1.111000			0.200	2.000021	.10,7021

¹⁸ See State v. Frampton, 95 Wn.2d 469, 627 P.2d 922 (1981) and State v. Martin, 94 Wn.2d 1, 614 P.2d 164 (1980).

MODIFIED TABLE	7 FR	OM UPDATED	REPORT,	KEEPING	PLEA,	DROPPING	VICTIM	HELD	HOSTAGE
logit DP_Senten Vics_1Total Plea	ce D a_Gu	_RaceB Appl ilty, level	iedAggCi (90) ;	.r_Num Li	nTotMi	tCircum De	efenses_	_Num l	nPriors
Iteration 0:	log	likelihood	= -53.13	8727					
Iteration 1:	log	likelihood	= -41.93	0564					
Iteration 2:	log	likelihood	= -41.62	2847					
Iteration 3:	log	likelihood	= -41.62	1278					
Iteration 4:	log	likelihood	= -41.62	1278					
Logistic regres	sion				N	umber of d	obs =		78
					L	R chi2(7)	=		23.03
					P	rob > chi2	2 =	0	.0017
Log likelihood :	= -4	1.621278			P	seudo R2	=	0	.2167
DP_Senten	 ce	Coef.	Std.	Err.	 Z	P> z	[90%]	Conf.	Interval]
D Race	+ eB	1.60189	.7731	.817	2.07	0.038	. 3301	 193	2.873661
AppliedAggCir N	um	.5795735	.251	185	2.31	0.021	.166	5411	.992736
LnTotMitCirc	um	1924647	.1377	282 -	-1.40	0.162	4190	075	.0340781
Defenses N	um	8127501	. 3912	228	-2.08	0.038	-1.456	5254	1692459
lnPrio	rs	011408	.1311	572 .	-0.09	0.931	2271	424	.2043265
Vics 1Tota	al I	7376041	.5794	988 -	-1.27	0.203	-1.690)795	.2155867
Plea Guil	ty İ	2944573	.7692	464 -	-0.38	0.702	-1.559	9755	.9708405
CO:	ns	7672557	.7915	007 -	-0.97	0.332	-2.069	9158	.5346471

MODIFIED TABLE 7 F	ROM UPDATED RE	EPORT, KEEPIN	NG PLEA,	DROPPING	PRIORS	
logit DP_Sentence :	D_RaceB Applie	edAggCir_Num	LnTotMit	Circum De	fenses_Num	Vics_1Total
Plea_Guilty Vics_A	nyHostage, lev	vel(90) ;				
Iteration 0: log	likelihood =	-53.99031				
Iteration 1: log	likelihood =	-41.175559				
Iteration 2: log	likelihood =	-40.751979				
Iteration 3: log	likelihood =	-40.749699				
Iteration 4: log	likelihood =	-40.749699				
Logistic regressio:	n		Nu	umber of o	obs =	79
			LI	R chi2(7)	=	26.48
			Pi	rob > chi2	=	0.0004
Log likelihood = $-$	40.749699		Ps	seudo R2	=	0.2452
DP_Sentence	Coef.	Std. Err.	 Z	P> z	[90% Conf	. Interval]
D_RaceB	+ 1.446545	.7843018	1.84	0.065	.1564829	2.736606
AppliedAggCir_Num	.582	.2489025	2.34	0.019	.1725918	.9914082
LnTotMitCircum		.1484752	-1.72	0.086	4989414	0105014
Defenses_Num	8267372	.3896202	-2.12	0.034	-1.46/605	1858689
vics_lTotal	8447283	.5//2446	-1.46	0.143	-1./94211	.1047545
Plea_Guilty	2936731	./919355	-0.37	0.711	-1.596291	1.008945
Vics_AnyHostage	1.7593045	.5/62541	1.32	0.188	1885491	1.707158
cons	8880676	.8038664	-1.10	0.269	-2.21031	.4341751

Interrogatory 32

Do you maintain that the results of any of the models presented in the *Updated Report* provide a basis to determine the percentage of variation in outcomes that is explained by the case characteristics included in the models? If yes, please identify the models and explain.

No, we do not maintain that the models presented in the *Updated Report* provide a basis to determine the percentage of explained variation. Logistic regression does not have an equivalent to the R-squared that is found in OLS regression. However, many people have tried to develop one, and there are a wide variety of pseudo-R-square statistics. Although Pseudo R-squared statistics cannot be interpreted independently or compared across datasets, they are valid and useful in evaluating multiple models predicting the same outcome using the same dataset.¹⁹ For these reasons, and because this statistic does not mean what R-square means in OLS regression (the proportion of variance explained by the predictors), we suggest using this statistic only to compare models using the same dataset. In this situation, the higher pseudo R-squared identifies the model that better predicts the outcome.

¹⁹ Scott J. Long, and Jeremy Freese, REGRESSION MODELS FOR CATEGORICAL DEPENDENT VARIABLES USING STATA (2nd Ed. College Station, Texas: StataCorp LP, 2006) pp.107-113.

The Statistical Consulting Group at UCLA point to the McKelvey and Zavoina's Pseudo R-squared measure as the closest to approximating an OLS R-squared that provides the percentage of variation explained:

"Attempts have been made to assess the accuracy of various pseudo R-squareds by predicting a continuous latent variable through OLS regression and its observed binary variable through logistic regression and comparing the pseudo R-squareds to the OLS R-squared. In such simulations, McKelvey & Zavoina's was the closest to the OLS R-squared."²⁰

In Table 11 below we present the results of multiple Pseudo R-squared values, comparing these values across three models: 1) the null model with no predictors, 2) the model with only legally relevant case characteristics (6 predictors), and 3) the same legal model with defendant race added (7 predictors.) We have highlighted the McKelbey and Zavoina Pseudo R-squared values for each model.

²⁰ Commonly Encountered Pseudo R-squareds. UCLA: Statistical Consulting Group, from https://stats.idre.ucla.edu/other/mult-pkg/faq/general/faq-what-are-pseudo-r-squareds/ (accessed July 7, 2017).

Table 11. Comparing Pseudo	o R-squared Val	ues Across Three Models					
Null Model (No pred	dictors)	Legally Relevant Charact	eristics (6	Legally Relevant Charact	eristics and		
	•	Predictors)		Defendant Race (7 pre	Detendant Race (7 predictors)		
Code: logit DP_Sentence		Code: logit DP_Sentence lnPriors AppliedAggCir_Num LnTotMi Defenses_Num Vics_AnyHost	s Vics_1Total tCircum cage	Code: logit DP_Sentence lnPriors Vics_1Total AppliedAggCir_Num LnTotMitCircum Defenses_Num Vics AnyHostage D RaceB			
	logit		logit		logit		
Log-likelihood Model Intercept-only	-55.957 -55.957	Log-likelihood Model Intercept-only	-42.830 -53.139	Log-likelihood Model Intercept-only	-40.166 -52.584		
Chi-square		Chi-square		Chi-square			
Deviance (df=81) LR (df=0)	111.914 0.000	Deviance (df=71) LR (df=6) p-value	85.659 20.618 0.002	Deviance (df=69) LR (df=7) p-value	80.333 24.835 0.001		
P2		P2		R2			
McFadden (adjusted)	0.000 -0.018	McFadden (adjusted)	0.194 0.062	McFadden McFadden (adjusted)	0.236 0.084		
McKelvey & Zavoina	0.000	McKelvey & Zavoina	0.439	McKelvey & Zavoina	0.496		
Cox-Snell/ML Cragg-Uhler/Nagelkerke	0.000 0.000	Cox-Snell/ML Cragg-Uhler/Nagelkerke	0.232 0.312	Cox-Snell/ML Cragg-Uhler/Nagelkerke	0.276 0.370		
Efron	0.000	Efron	0.226	Efron	0.267		
Tjur's D	0.000	Tjur's D	0.229	Tjur's D	0.274		
Count (adjusted)	-0.000	Count (adjusted)	0.654	Count (adjusted)	0.242		
IC		IC		IC			
AIC	113.914	AIC	99.659	AIC	96.333		
AIC divided by N BIC (df=1)	1.389 116.320	AIC divided by N BIC (df=7)	1.278 116.156	AIC divided by N BIC (df=8)	1.251 115.083		
Variance of	+ 	Variance of		Variance of			
e y-star	3.290 3.290	e y-star	3.290 5.867	e y-star	3.290 6.533		

All Pseudo R-squared measures presented above show that the model containing case characteristics is a better fit of the data than a model containing no predictors (called the null model.) These same measures also demonstrate that including defendant race when controlling for relevant legal characteristics improves the model fit. The McKelvey & Zavoina measure for the model containing only legally relevant case characteristics is 0.44; when defendant race is included, this measure increases to .50. Although none of the Pseudo R-squared measures can be interpreted as an exact percentage of variation explained, none of these values approaches 1.0, indicating there is much unexplained variation in the decision to impose death.

IN THE SUPREME COURT OF THE STATE OF WASHINGTON

STATE OF WASHINGTON,

Respondent,

٧.

NO. 88086-7

ALLEN GREGORY,

Appellant.

DECLARATION OF DOCUMENT FILING AND SERVICE

I, MARIA ARRANZA RILEY, STATE THAT ON THE 12^{TH} DAY OF JULY, 2017, I CAUSED THE ORIGINAL **RESPONSE TO COMMISSIONER'S INTERROGATORIES TO PARTIES' EXPERTS** TO BE FILED IN THE **WASHINGTON STATE SUPREME COURT** AND A TRUE COPY OF THE SAME TO BE SERVED ON THE FOLLOWING IN THE MANNER INDICATED BELOW:

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[talner@aclu-wa.org]	(X)	E-SERVICE VIA PORTAL
[X] MARC SHAPIRO	()	U.S. MAIL
ORRICK HERRINGTON & SUTCLIFFE LLP	()	HAND DELIVERY
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		Washington Appellate Project 701 Melbourne Tower 1511 Third Avenue Seattle, WA 98101 🕾 (206) 587-2711

[X] ARAVIND SWAMINATHAN	()	U.S. MAIL
ORRICK HERRINGTON & SUTCLIFFE LLP	()	HAND DELIVERY
[aravind@orrick.com]	(X)	E-SERVICE VIA PORTAL
[X] JOHN WOLFE	()	U.S. MAIL
ORRICK HERRINGTON & SUTCLIFFE LLP	()	HAND DELIVERY
[john.wolfe@orrick.com]	(X)	E-SERVICE VIA PORTAL
[X] CASSANDRA STUBBS	()	U.S. MAIL
ACLU	()	HAND DELIVERY
[cstubbs@aclu.org]	(X)	E-SERVICE VIA PORTAL
[X] JEFFERY ROBINSON	()	U.S. MAIL
ACLU	()	HAND DELIVERY
[robinson@sgb-law.com]	(X)	E-SERVICE VIA PORTAL

SIGNED IN SEATTLE, WASHINGTON THIS 12TH DAY OF JULY, 2017.

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