

# **Measuring the Impact of a Lower BAC in Utah**

**A Frequentist and Bayesian Comparison**

# Introduction

**0.05 BAC impact is measured by DUI related collisions**

- **What:**

- Utah lowered it's BAC limit on 2018-12-30 from 0.08% to 0.05%

- **Expectations:**

- The number alcohol related collisions and collisions involving death will decrease.

- **How to measure:**

- We look at weekly DUI related collisions in Utah from 2015-01-01 to 2019-03-31

# Methodology

## We compare Frequentist and Bayesian Methodologies

### - Frequentist

1. Identify plausible ARIMA model on pre-intervention data
2. Run model on full series with a treatment binary included
3. Ensure quality measures hold (i.e. residual checks, significance, convergence)
4. Evaluate treatment binary

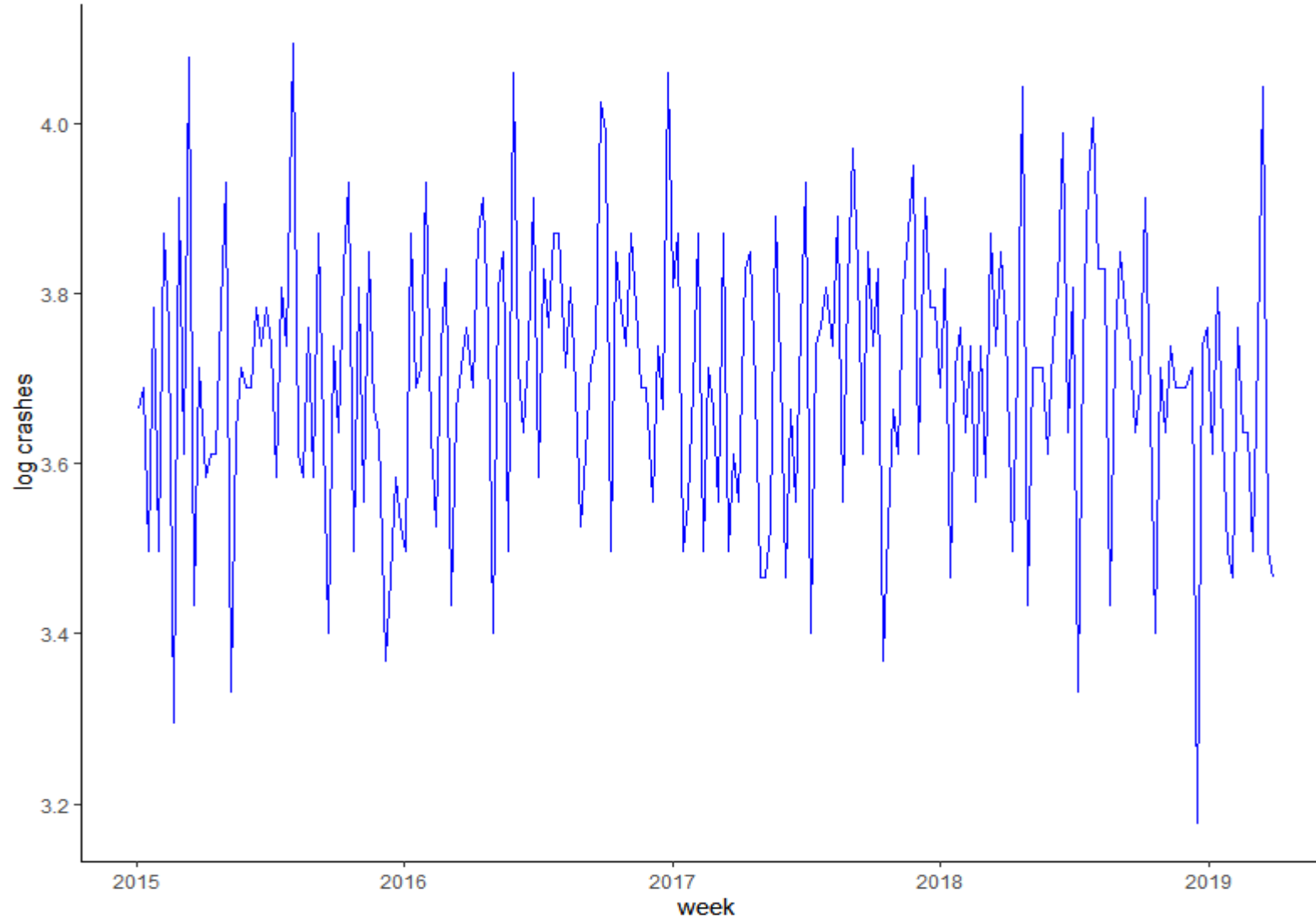
### - Bayesian

1. Build several state-space models on the full dataset and evaluate on:
  - RMSE, RMSE on a holdout, R-Square, Harvey's GOF, Posterior Predictive Checks
2. Include treatment binary and evaluate impact based on:
  - Posterior Inclusion Probability
  - HPD interval given inclusion
3. **Robustness Check:** Pre-built Bayesian State-Space Causal Impact Analysis

# Frequentist Approach: Unit Root Testing

Perform visual inspection as first step of Unit Root Tests

Log Weekly DUI Collisions



- No obvious deterministic trend
- Stationary test to include drift
- Seasonality potentially exists

# Frequentist Approach: Unit Root Testing

HEGY tests for unit roots at each seasonal frequency

- Seasonal Unit Root Example: SAR(1)

$$y_t = \alpha y_{t-52} + \varepsilon$$

$$\alpha = 1$$

- HEGY Test
  - $H_0$ : Unit Root
  - Test for a unit root at each seasonal frequency

# Frequentist Approach: Unit Root Testing

## HEGY statistics on level data show seasonal unit roots

- H<sub>0</sub>: Unit Root

### HEGY Test Statistics

test			test		
	statistic	p-value		statistic	p-value
t_1	-2.68	0.061	F_31:32	1.85	0.110
t_2	-1.16	0.236	F_33:34	2.24	0.073
F_3:4	1.82	0.113	F_35:36	2.09	0.085
F_5:6	3.30	0.025 *	F_37:38	1.92	0.102
F_7:8	1.46	0.169	F_39:40	1.57	0.151
F_9:10	2.64	0.048 *	F_41:42	4.89	0.005 **
F_11:12	4.18	0.009 **	F_43:44	5.53	0.003 **
F_13:14	1.93	0.102	F_45:46	3.84	0.014 **
F_15:16	0.34	0.582	F_47:48	1.91	0.103
F_17:18	2.81	0.041 *	F_49:50	1.94	0.101
F_19:20	2.81	0.041 *	F_51:52	0.41	0.531
F_21:22	0.83	0.331	F_2:52	3.67	0.253
F_23:24	1.13	0.243	F_1:52	3.78	0.000 ***
F_25:26	2.49	0.056			
F_27:28	0.42	0.525			
F_29:30	3.59	0.018 **			

Sig. at 0.05  
Insignificant



# Frequentist Approach: Unit Root Testing

## HEGY statistics on 1 seasonally difference show no unit roots

- H<sub>0</sub>: Unit Root

HEGY Test Statistics

	test			test	
	statistic	p-value		statistic	p-value
t_1	-2.68	0.061	F_31:32	1.85	0.110
t_2	-1.16	0.236	F_33:34	2.24	0.073
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F_25:26	2.49	0.056			
F_27:28	0.42	0.525			
F_29:30	3.59	0.018 **			

Sig. at 0.05  
Insignificant

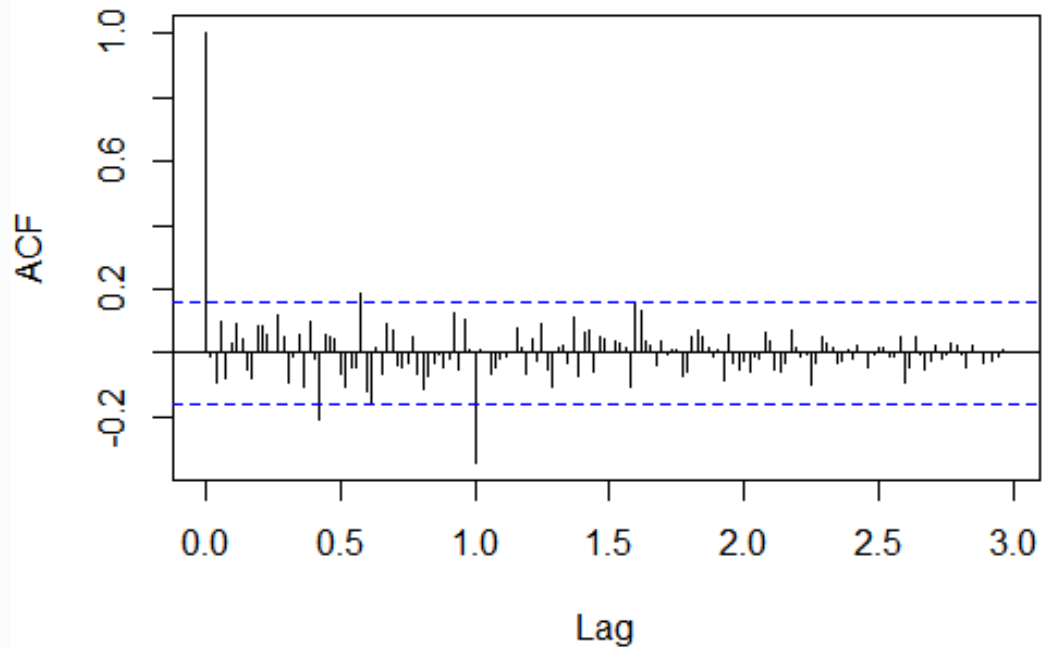
HEGY Test Statistics on 1 Seasonal Diff.

	test			test	
	statistic	p-value		statistic	p-value
t_1	-3.50	1.000	F_31:32	3.68	0.013 **
t_2	-1.18	0.991	F_33:34	3.72	0.013 **
F_3:4	3.55	0.013 **	F_35:36	3.22	0.014 **
F_5:6	4.45	0.012 **	F_37:38	4.56	0.011 **
F_7:8	0.62	0.018 **	F_39:40	3.67	0.013 **
F_9:10	2.53	0.015 **	F_41:42	6.59	0.000 ***
F_11:12	6.10	0.008 **	F_43:44	3.99	0.013 **
F_13:14	5.57	0.009 **	F_45:46	4.19	0.012 **
F_15:16	3.42	0.014 **	F_47:48	3.19	0.014 **
F_17:18	4.97	0.011 **	F_49:50	2.97	0.014 **
F_19:20	0.63	0.018 **	F_51:52	1.70	0.017 **
F_21:22	5.24	0.010 **	F_2:52	6.99	0.304
F_23:24	3.66	0.013 **	F_1:52	7.13	0.000 ***
F_25:26	5.11	0.010 **			
F_27:28	1.39	0.017 **			
F_29:30	0.99	0.017 **			

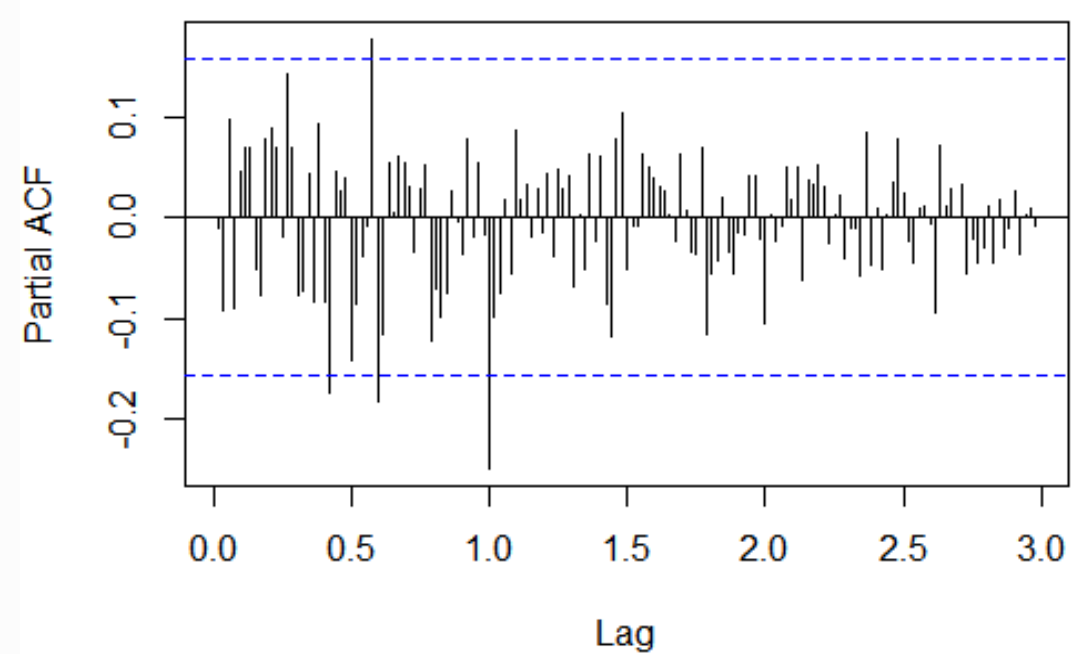
# Frequentist Approach: ARIMA Modeling

HEGY statistics on 1 seasonally difference show no unit roots

Seasonally Differenced ACF



Seasonally Differenced PACF

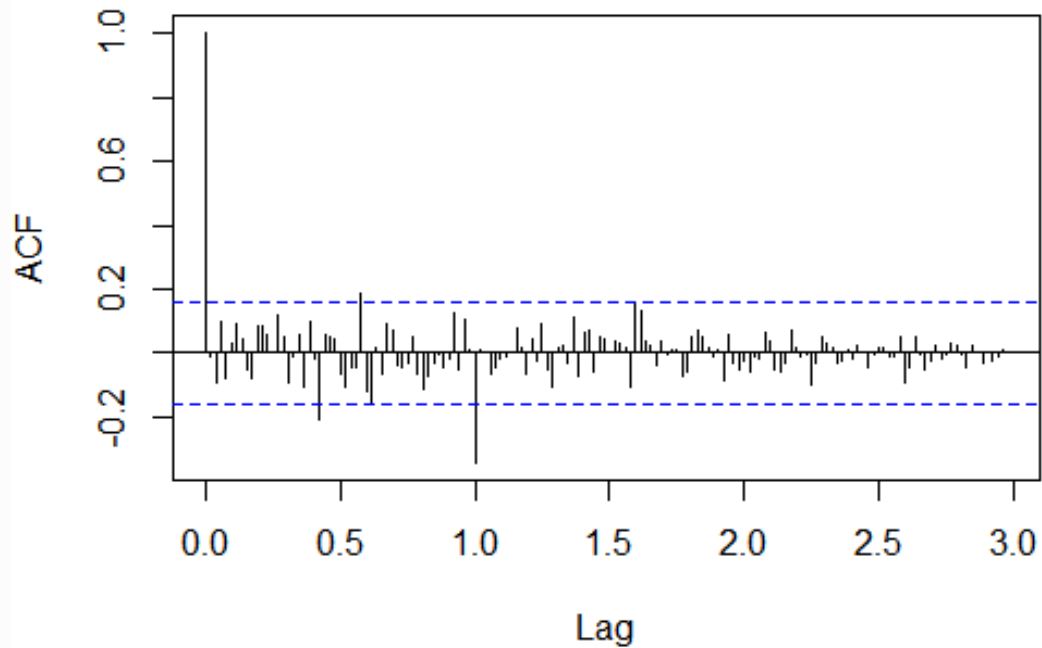




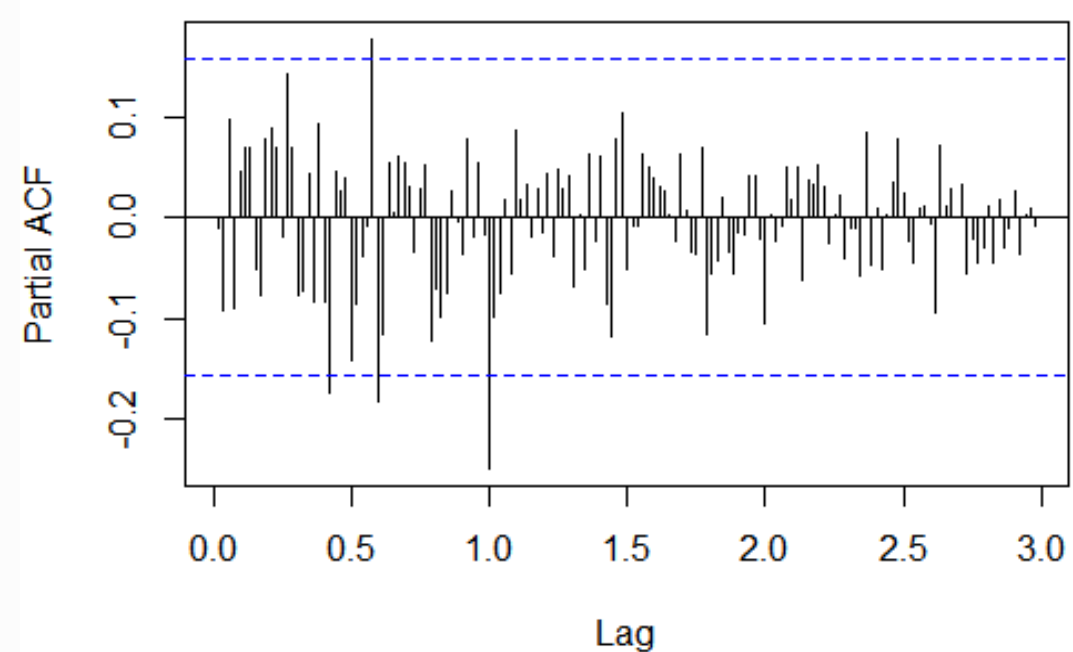
# Frequentist Approach: ARIMA Modeling

HEGY statistics on 1 seasonally difference show no unit roots

Seasonally Differenced ACF



Seasonally Differenced PACF



Move forward with a SAR(1), SMA(1), pure noise models  
(of seasonal order 1)

# Frequentist Approach: ARIMA Modeling

**SAR(1) is only model that passes coefficient quality checks.**

	SAR(1)	SMA(1)	Pure Noise	
Estimate (P-Value)	sar1: <b>-0.49 (0.000)</b> treatment: <b>-0.03 (0.449)</b>	sma1: <b>-0.99 (0.223)</b> treatment: <b>-0.04 (0.421)</b>	treatment: <b>-0.04 (0.46)</b>	} Coefficient Quality

\*\*Ljung-Box Test  
H<sub>0</sub>: Independent errors -- White Noise

# Frequentist Approach: ARIMA Modeling

**SAR(1) and SMA(1) have lowest AIC and BIC.**

	SAR(1)	SMA(1)	Pure Noise	
Estimate (P-Value)	sar1: <b>-0.49 (0.000)</b> treatment: <b>-0.03 (0.449)</b>	sma1: <b>-0.99 (0.223)</b> treatment: <b>-0.04 (0.421)</b>	treatment: <b>-0.04 (0.46)</b>	Coefficient Quality
AIC	<b>-60.87</b>	<b>-0.71</b>	<b>-0.31</b>	
BIC	<b>-51.47</b>	<b>-0.61</b>	<b>-0.25</b>	Model fit

\*\*Ljung-Box Test  
H<sub>0</sub>: Independent errors -- White Noise

# Frequentist Approach: ARIMA Modeling

## SAR(1) and SMA(1) pass Ljung-Box Test.

	SAR(1)	SMA(1)	Pure Noise	
Estimate (P-Value)	sar1: <b>-0.49 (0.000)</b> treatment: <b>-0.03 (0.449)</b>	sma1: <b>-0.99 (0.223)</b> treatment: <b>-0.04 (0.421)</b>	treatment: <b>-0.04 (0.46)</b>	Coefficient Quality
AIC	<b>-60.87</b>	<b>-0.71</b>	<b>-0.31</b>	Model fit
BIC	<b>-51.47</b>	<b>-0.61</b>	<b>-0.25</b>	
Ljung-Box Test				
lag 5 p-value	<b>0.41</b>	<b>0.21</b>	<b>0.08</b>	Residual Quality
lag 10 p-value	<b>0.43</b>	<b>0.48</b>	<b>0.17</b>	
lag 15 p-value	<b>0.13</b>	<b>0.2</b>	<b>0.11</b>	
lag 20 p-value	<b>0.15</b>	<b>0.34</b>	<b>0.05</b>	
lag 25 p-value	<b>0.11</b>	<b>0.31</b>	<b>0.03</b>	

\*\*Ljung-Box Test  
H<sub>0</sub>: Independent errors -- White Noise

# Frequentist Approach: ARIMA Modeling

Treatment variable shows no significant effect on collisions

SAR(1)

Estimate (P-Value)

sar1: -0.49 (0.000)  
treatment: -0.03 (0.449)

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Conclusion: No significant impact of lowering the BAC on DUI related collisions

# Bayesian Approach: Model Comparison

We compare 5 state-space models

**Model 1:** Local Linear Trend & Seasonality

**Model 2:** Seasonality only

**Model 3:** Local Linear Trend only

**Model 4:** Semi Local Linear Trend only

**Model 5:** Semi Local Linear Trend & Seasonality

Modeled using BSTS in R: <https://www.rdocumentation.org/packages/bsts/versions/0.9.0>

## Evaluation

- RMSE on one-step-ahead predictions
- RMSE on a holdout
- $R^2$  & Harvey's GOF (random walk  $R^2$ )
- Posterior Predictive Checks

# Bayesian Approach: Model Comparison

Models 4 & 5 lead other models across evaluation criteria

	R <sup>2</sup>	HarveyGOF	RMSE (one-step)	RMSE (holdout)
Model 1	11.7%	42.2%	0.181	0.172
Model 2	9.1%	48.3%	0.171	4.497
Model 3	-0.6%	48.5%	0.171	0.173
Model 4	5.6%	49.9%	0.169	0.168
Model 5	16.5%	43.5%	0.179	0.162

# Bayesian Approach: Model Comparison

Models 4 & 5 lead other models across evaluation criteria

	R <sup>2</sup>	HarveyGOF	RMSE (one-step)	RMSE (holdout)	
Model 1	11.7%	42.2%	0.181	0.172	
Model 2	9.1%	48.3%	0.171	4.497	
Model 3	-0.6%	48.5%	0.171	0.173	
Model 4	5.6%	49.9%	0.169	0.168	Semi Local Linear Trend
Model 5	16.5%	43.5%	0.179	0.162	Semi Local Linear Trend & Seasonality

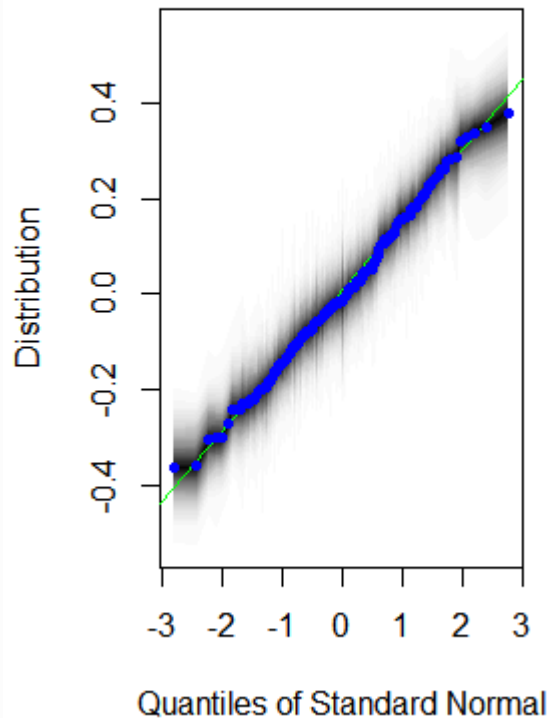


# Bayesian Approach: PPC Residual Checks

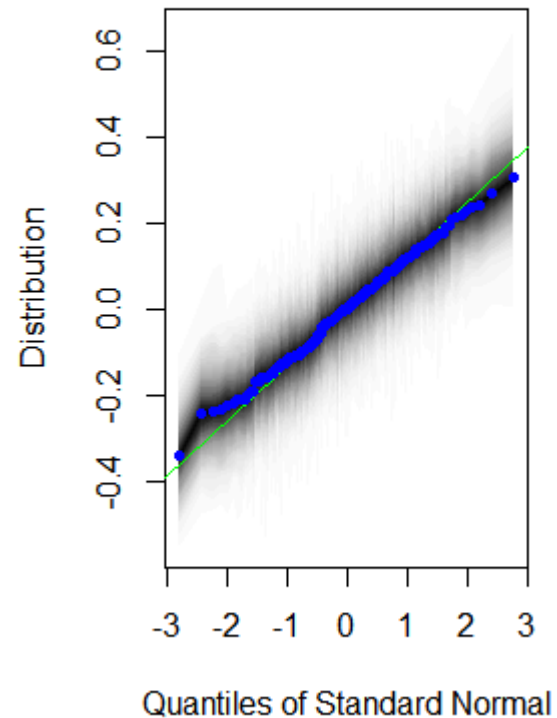
Q-Q plots show both model residuals are approx. normal

## Normal Q-Q Plots

Model 4



Model 5

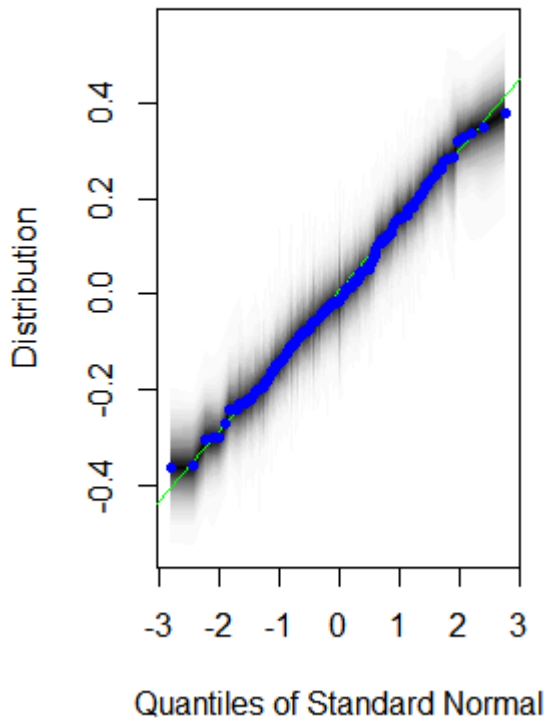


# Bayesian Approach: PPC Residual Checks

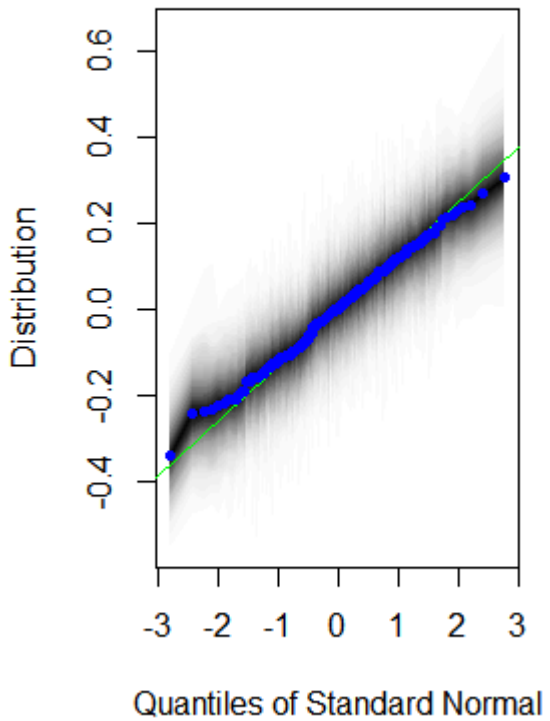
Neither model has a problem with serial correlation.

Normal Q-Q Plots

Model 4

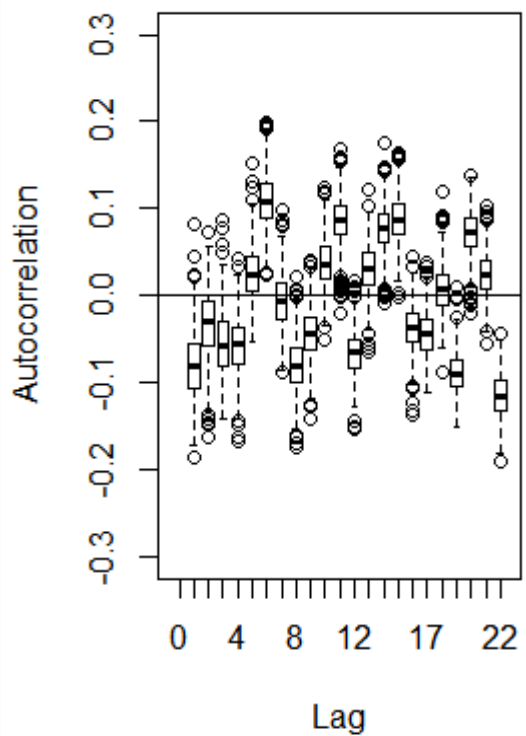


Model 5

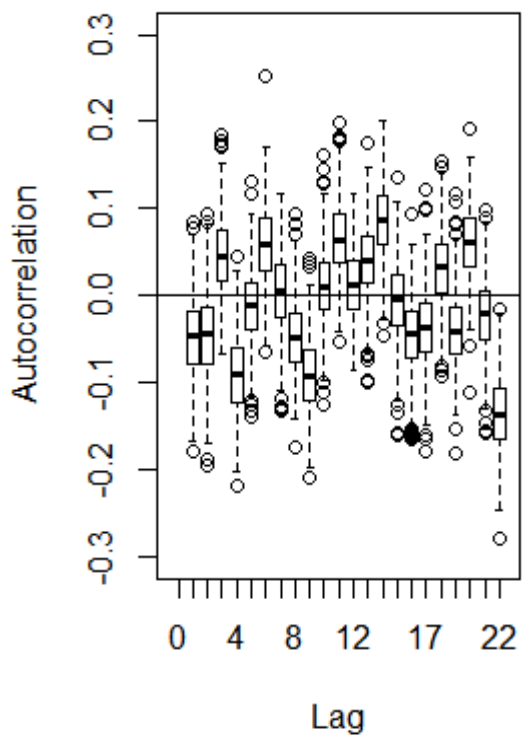


Residual ACF

Model 4



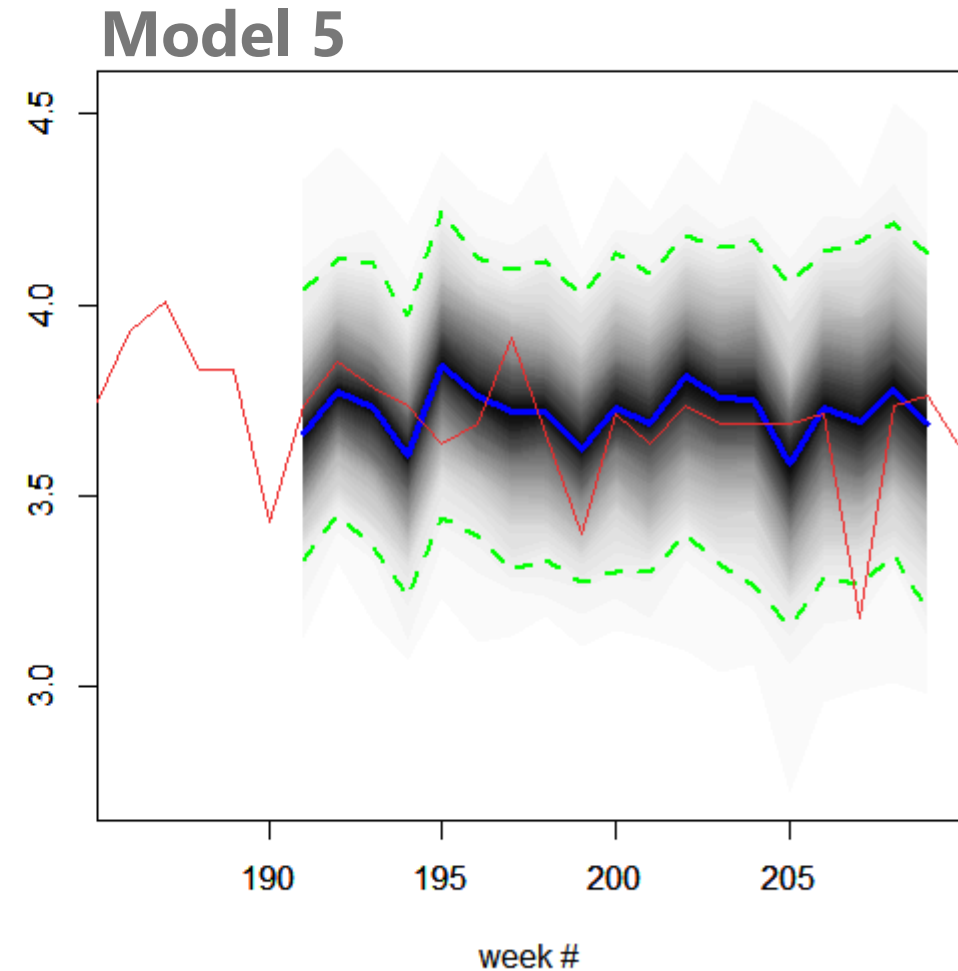
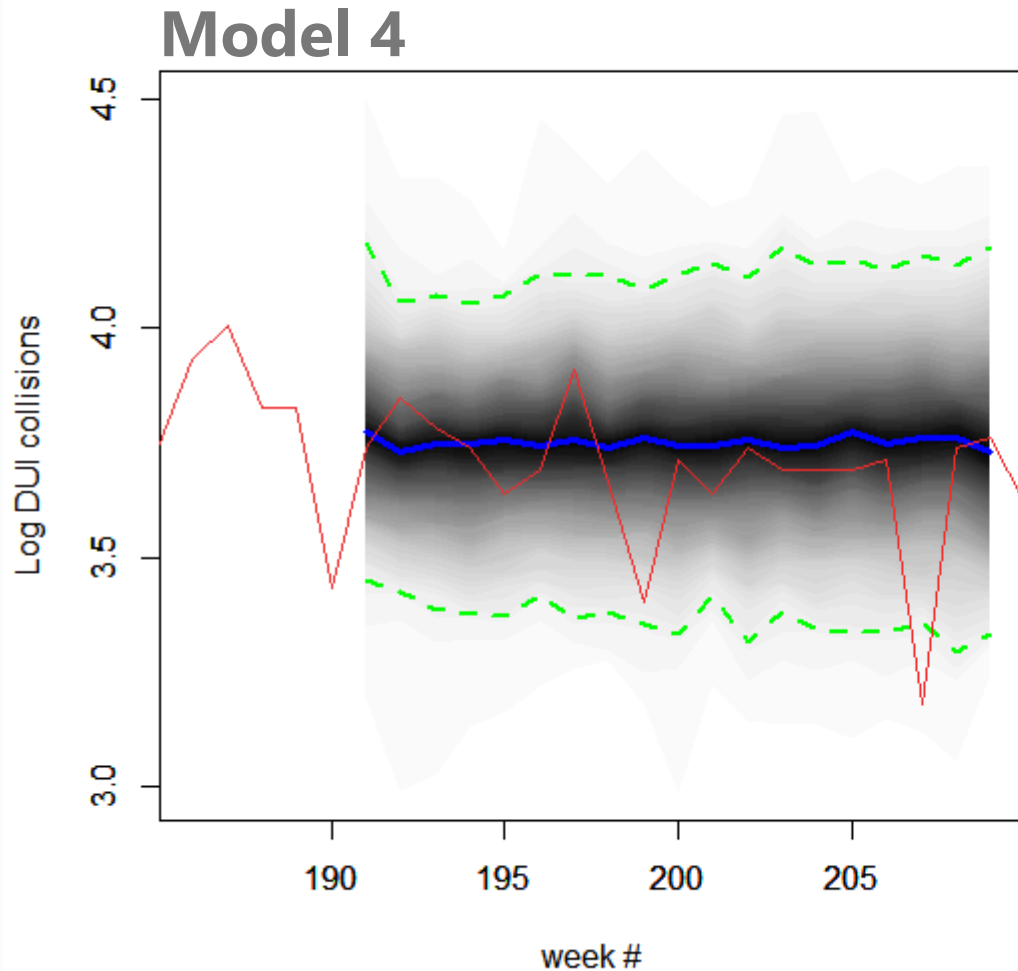
Model 5



# Bayesian Approach: Graphical Predictive Checks

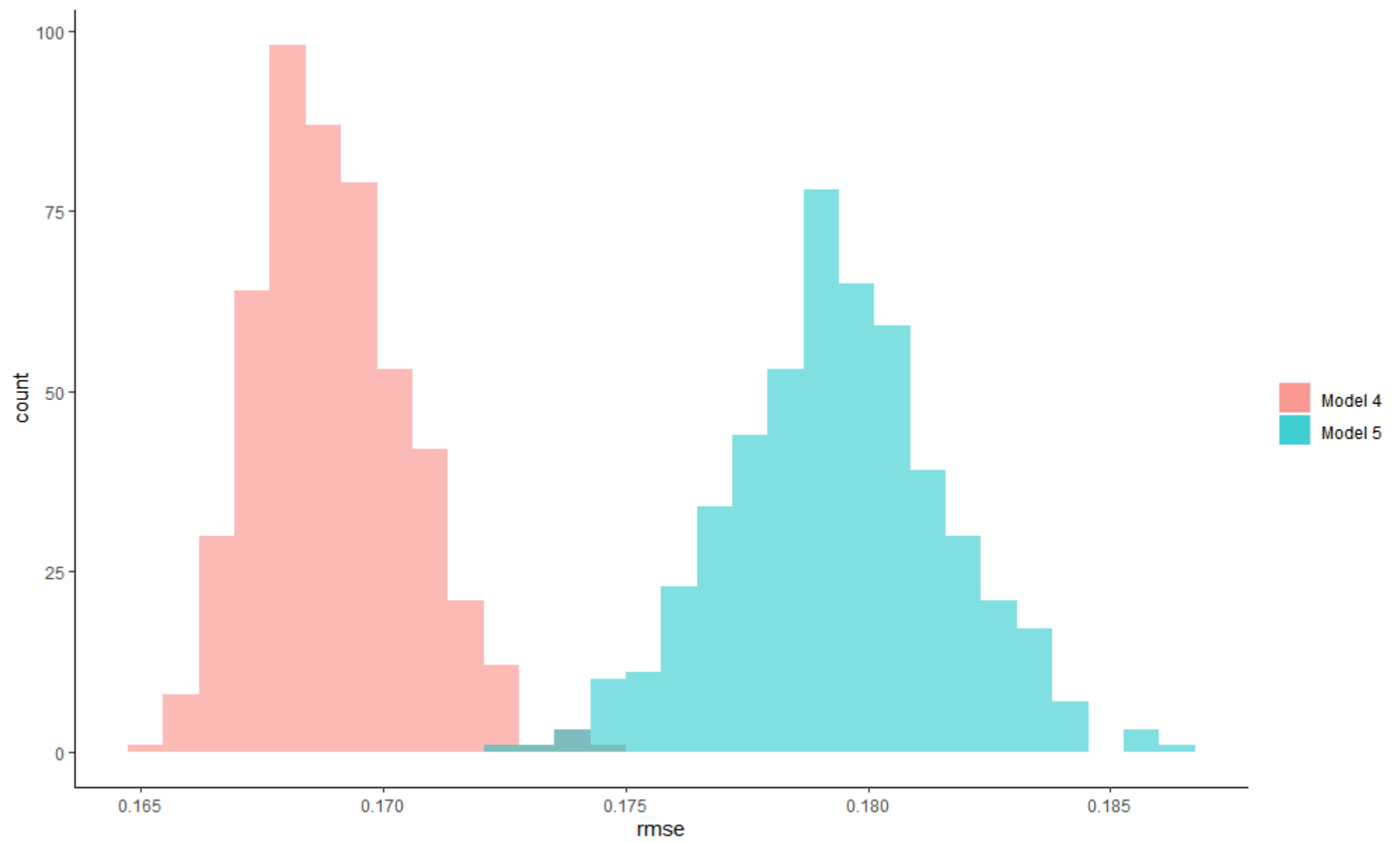
Model 5 appears to move closely with observed data

## Poster Predictions (20-step-ahead)



# Bayesian Approach: Graphical Predictive Checks

## Model 4 wins in terms of one-step-ahead RMSE



# Bayesian Approach: Treatment Evaluation

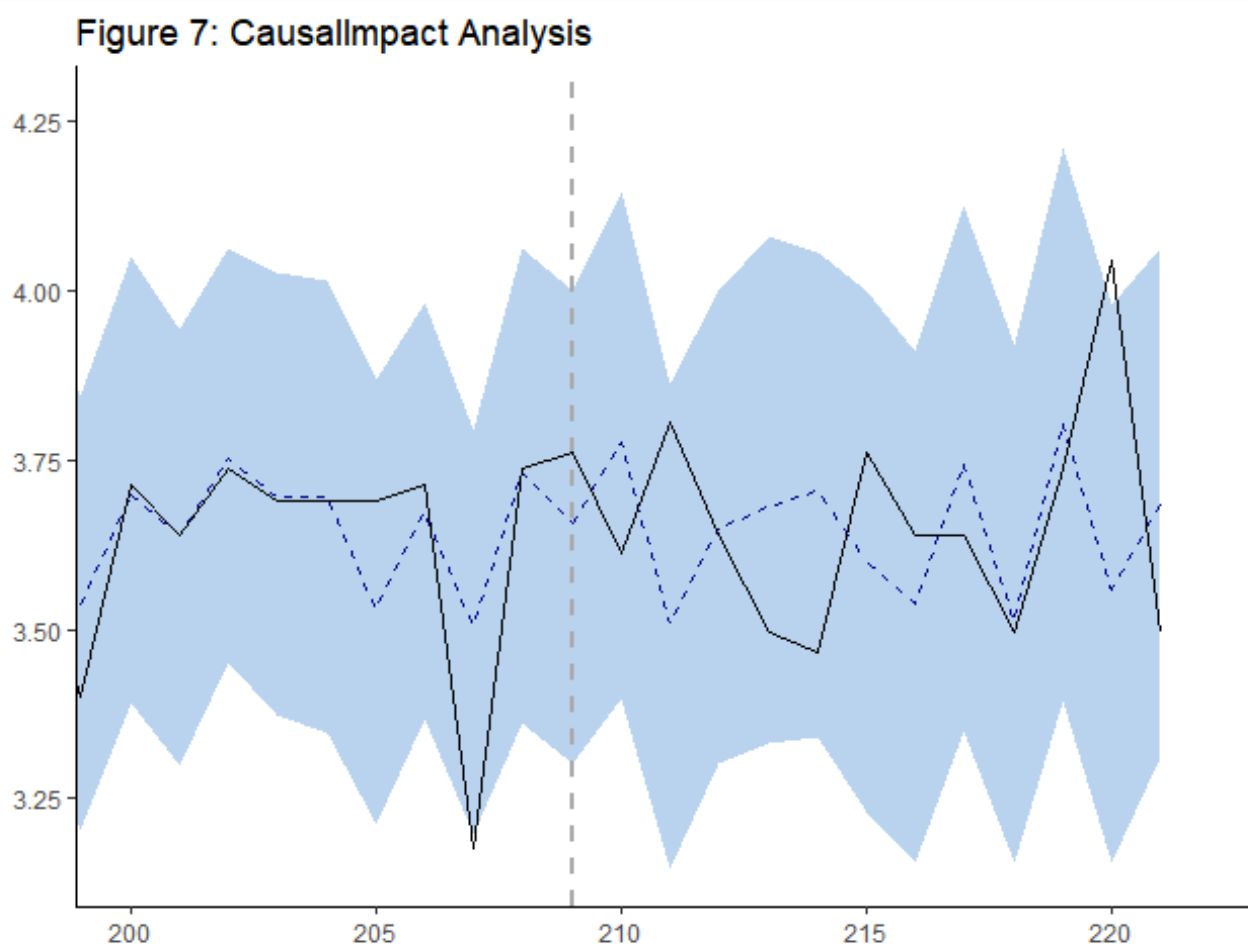
Neither model shows a significant effect of a lower BAC

	Inclusion Probability	Coefficient (95% HPD)
Model 4	<b>1.4%</b>	<b>0.045 (-0.08, 0.17)</b>
Model 5	<b>1.2%</b>	<b>0.036 (-0.18, 0.25)</b>

- Both models show inclusion probability much lower than conventional 10% threshold
- 95% HPD for both models includes zero meaning we can't determine a significant effect.

# Robustness Check

Robustness check also shows zero effect of lower BAC.



Expected log DUI collisions: **3.65 (3.35, 3.96)**  
Observed log DUI collisions: **3.65**  
Effect size: **0.0001 (-0.3, 0.3)**

Built in state-space causal modeling using Causallmpact R package provides a robustness check to previous results. Note, this is an entirely different methodology using our same state-space Model 5.

# Conclusion

- **Results:**
  - Both Frequentist and Bayesian methods show no significant impact
- **Limitations:**
  - DUI related deaths is what most previous research has studied
  - It's customary to normalize by vehicle miles driven.
  - Data only includes up to the first quarter of 2019