

# **Almond Prices, Production, and California Weather.** (1980 to 2021)

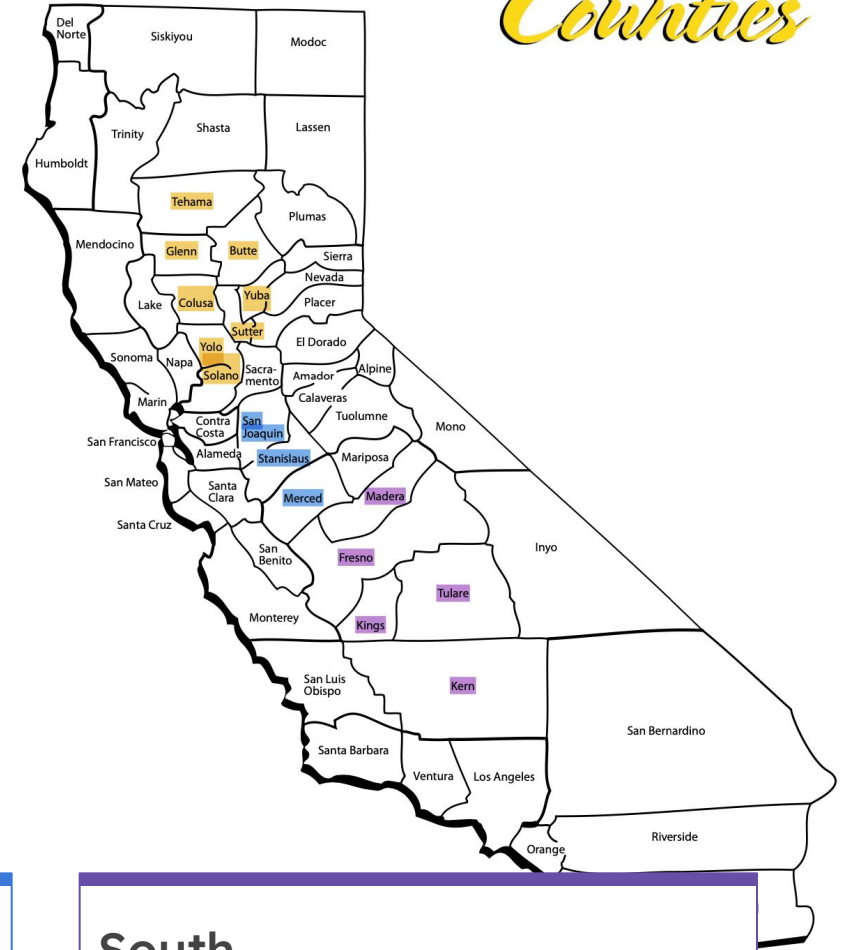
Doni Obidov & Joe Mallonee



# California

CALIFORNIA  
*Counties*

- ***80% of the world's almond supply.***
- Exported \$4.9B worth in 2019.
- Concentrated geography (16 counties) provides an ideal context for analysis.



## North

Butte, Colusa, Glenn, Solano, Sutter, Tehama, Yolo, Yuba

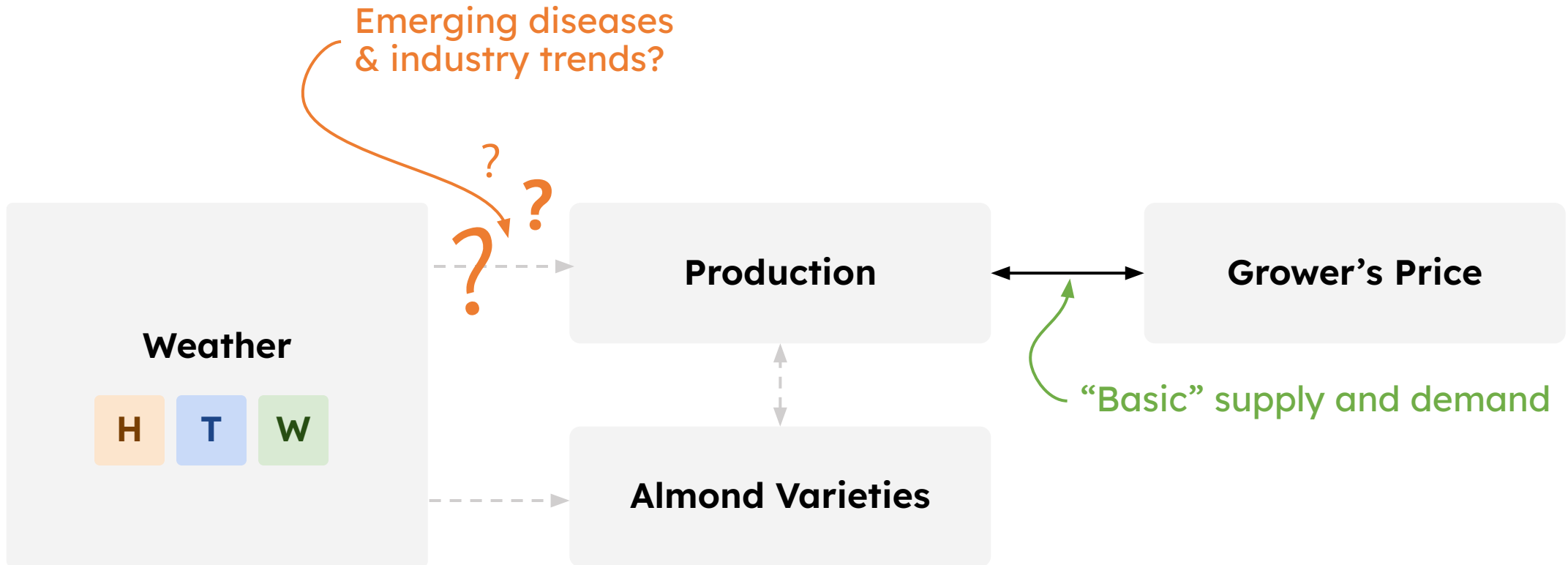
## Central

Merced, San Joaquin, Stanislaus

## South

Fresno, Kern, Kings, Madera, Tulare

# Research Goals



# Research Goals

You *might* assume: weather conditions influence the production and price of almonds. **Is this true?**



Can we identify factors that *consistently* improve BIC?

(we're not forecasting!)

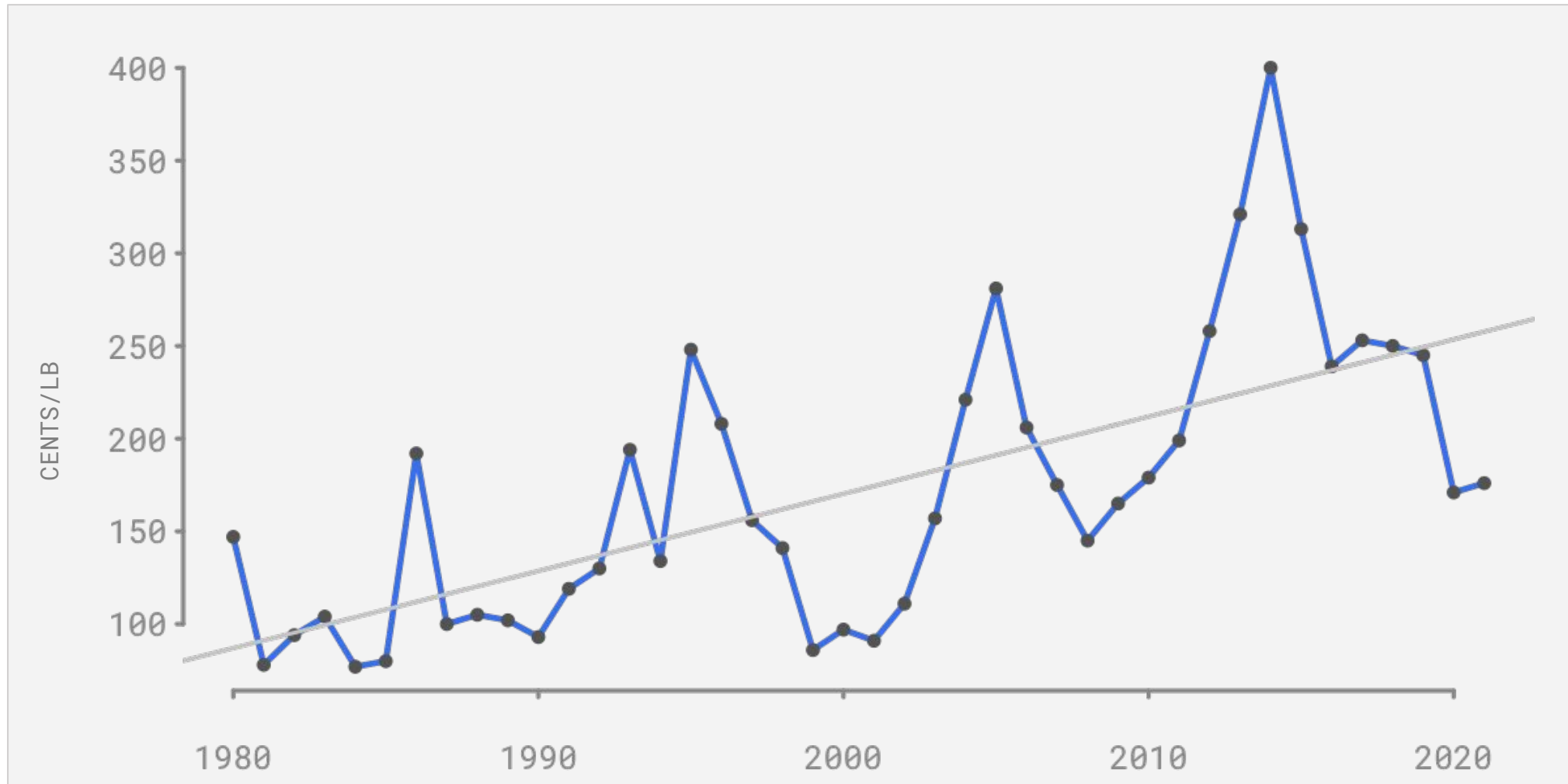


Can regressors/covariates lead to more parsimonious models?

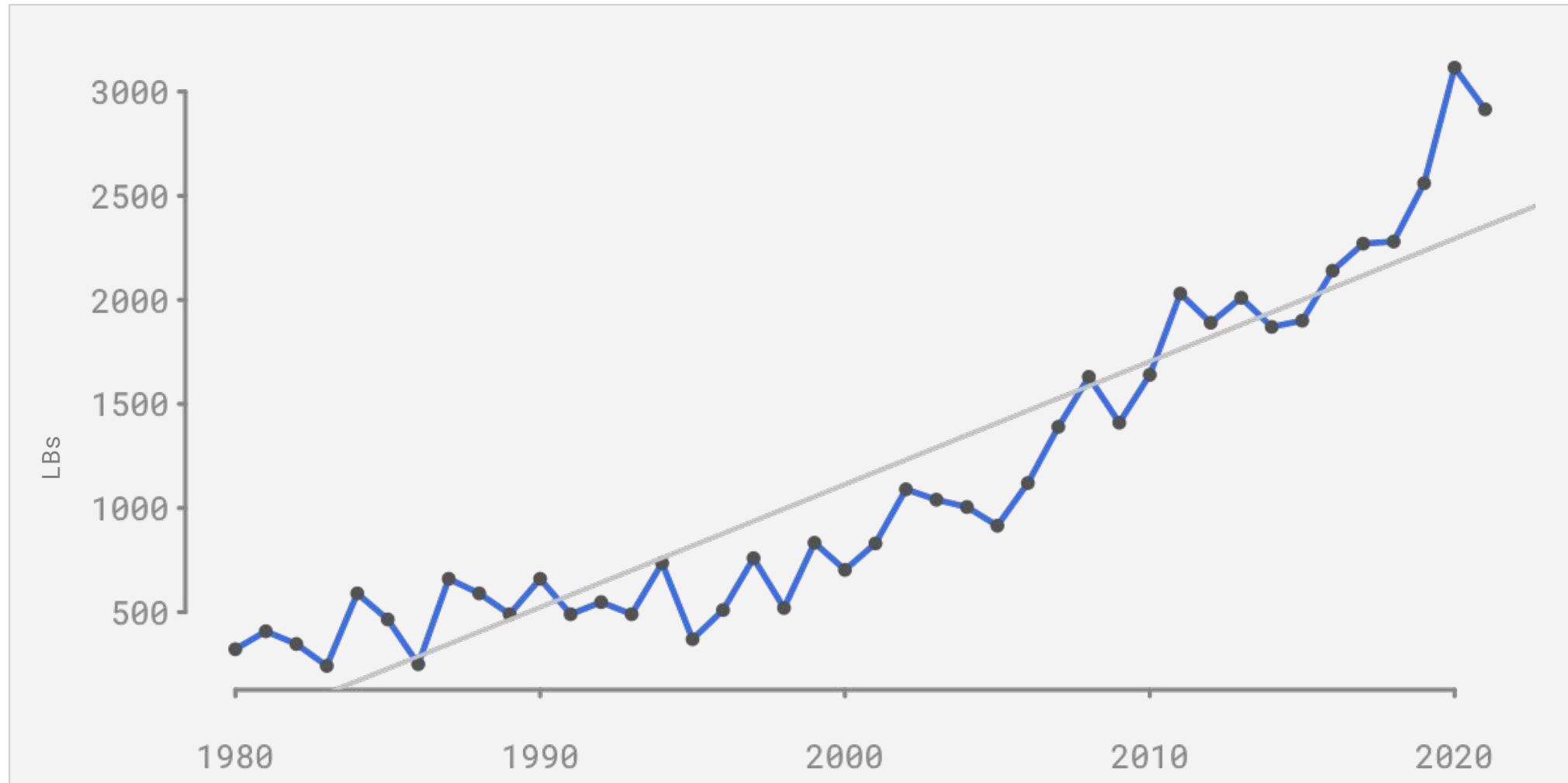
(simpler models with equal BIC performance?!)



# Price of California Almonds, 1980 to 2021

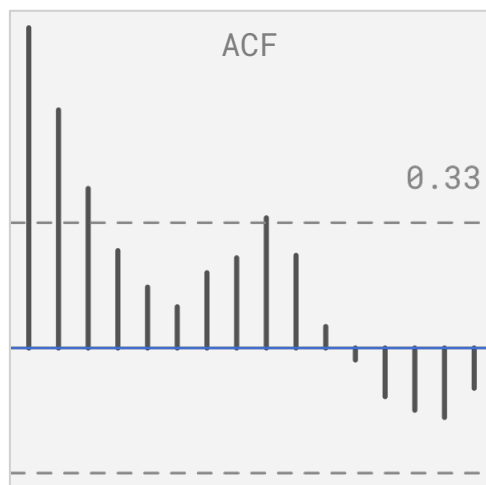
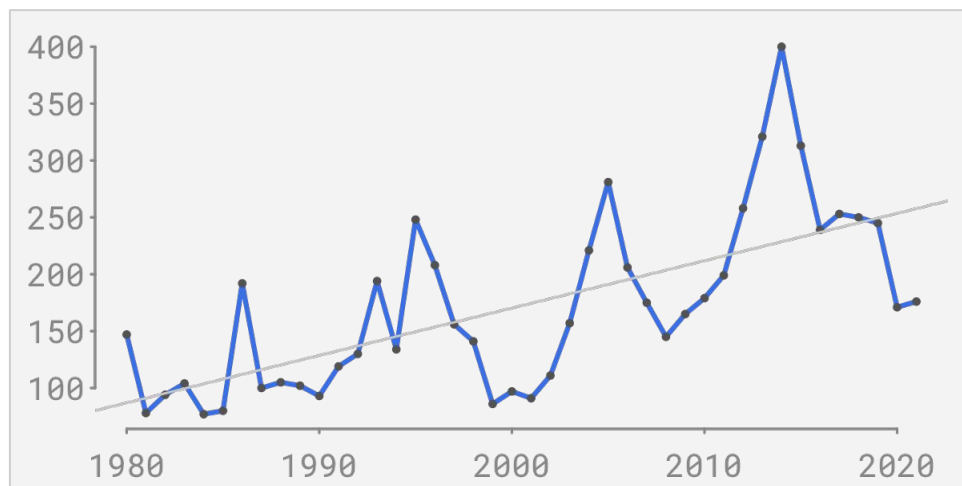


# Production of California Almonds, 1980 to 2021





# Price

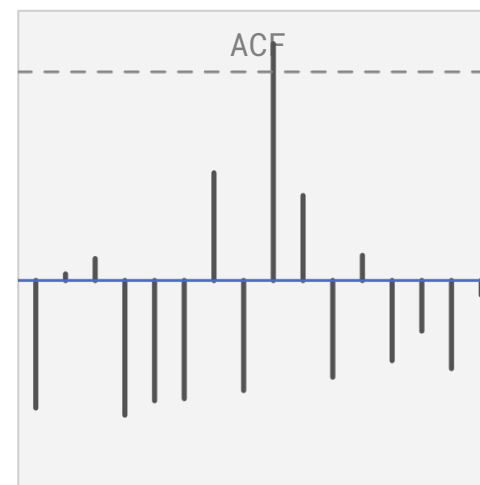
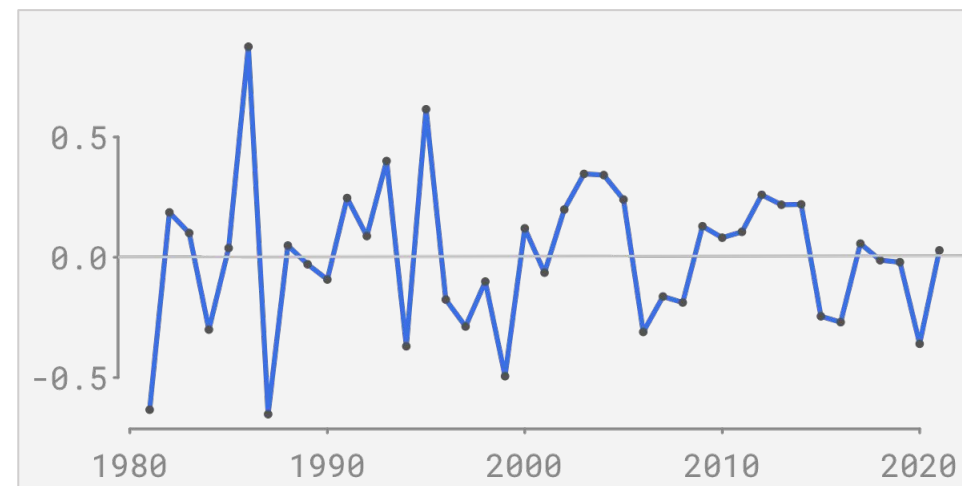


? **ADF** 0.049

? **PP** 0.045

x **KPSS** 0.013

# diff log Price



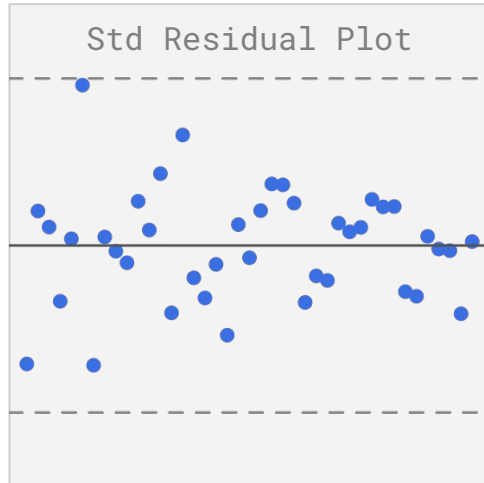
✓ **ADF** 0.038

✓ **PP** 0.010

✓ **KPSS** 0.100

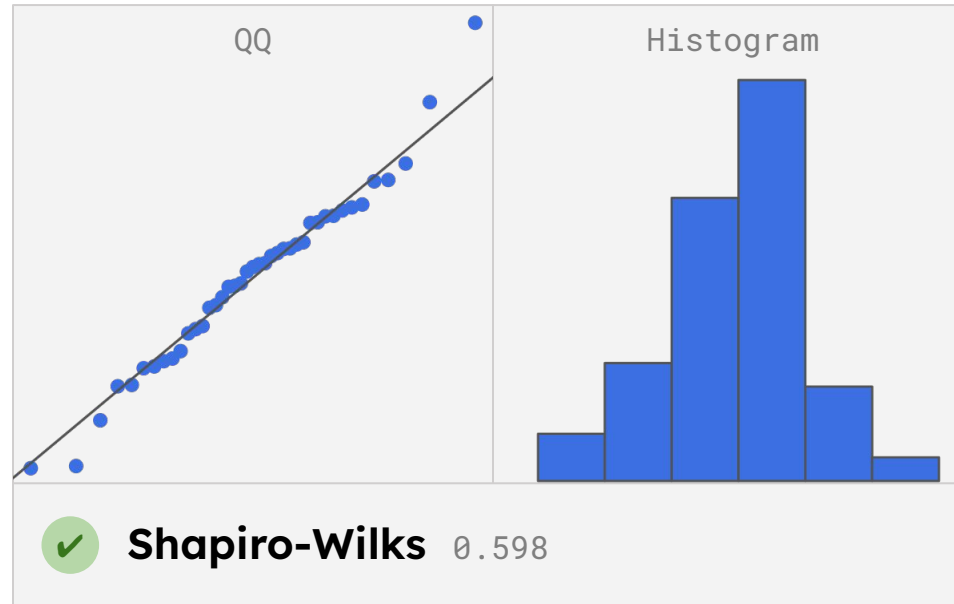


# diff log Price, Residual Analysis

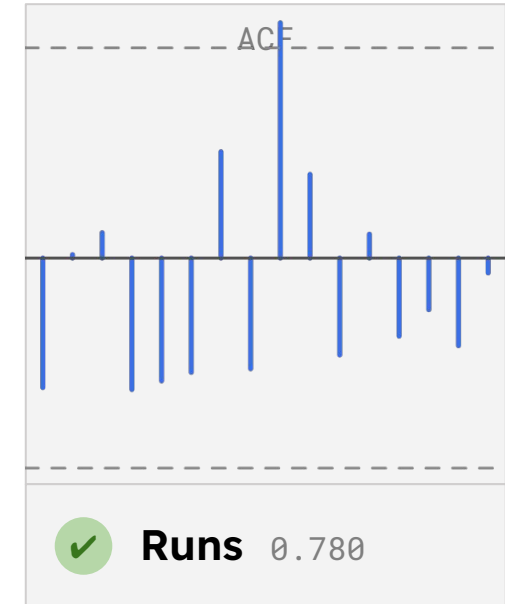


✓ Zero-Mean

? Homoscedasticity

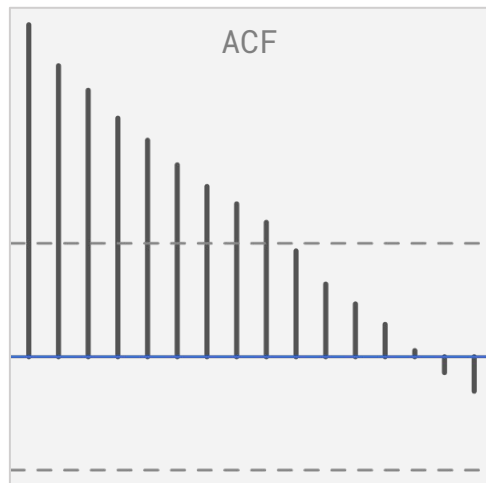
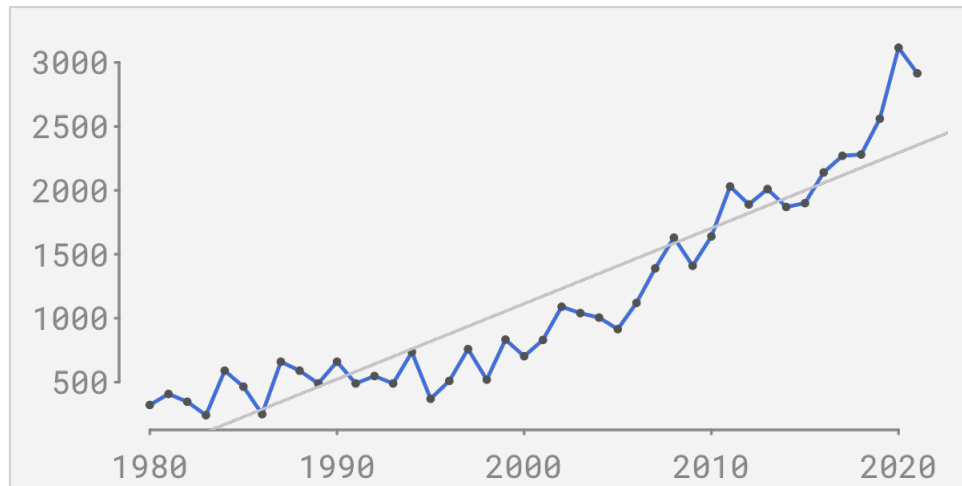


✓ Normality



✓ Independence

# Production

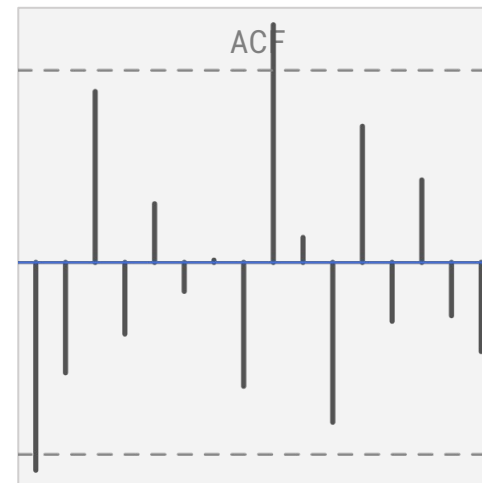
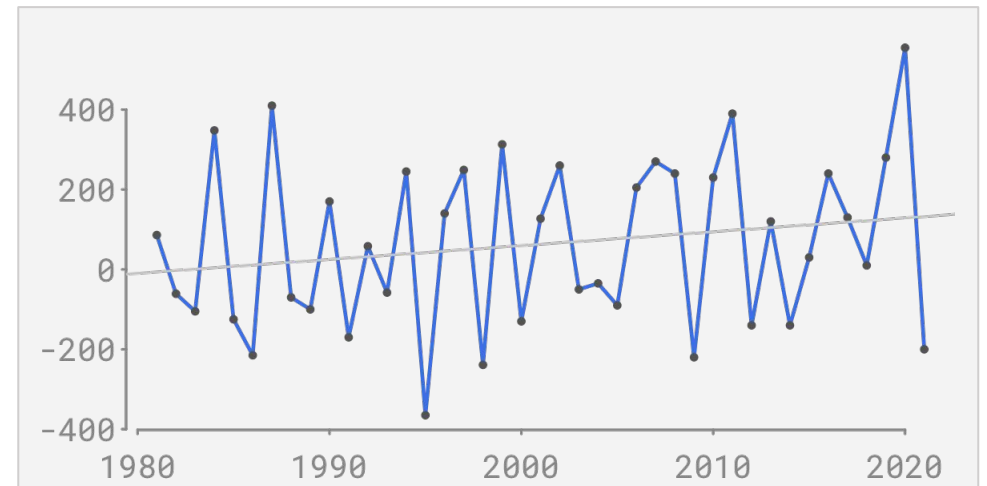


**✗ ADF** 0.990

**✗ PP** 0.704

**✗ KPSS** 0.010

# diff Production

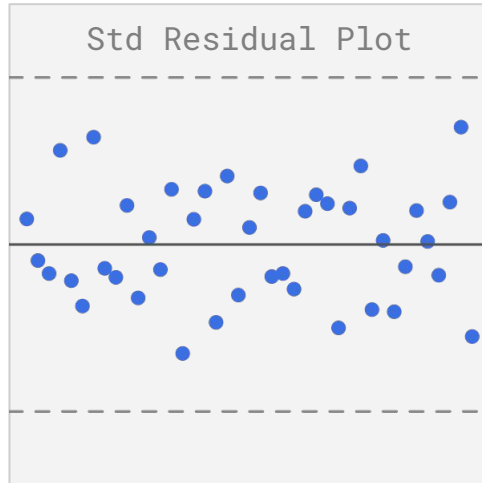


**✓ ADF** 0.016

**✓ PP** 0.010

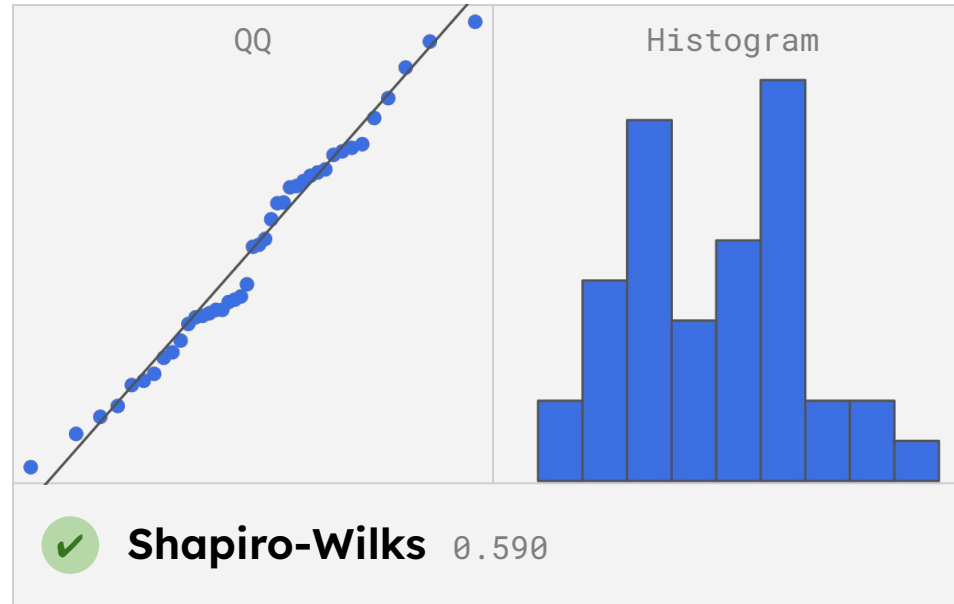
**✓ KPSS** 0.083

# diff Production, Residual Analysis

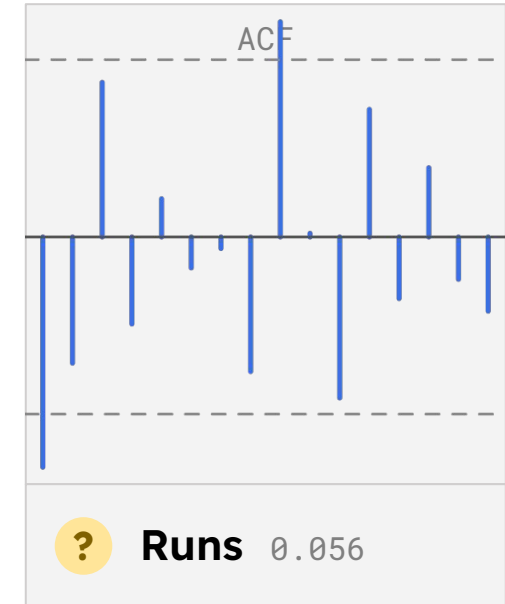


✓ Zero-Mean

✓ Homoscedasticity

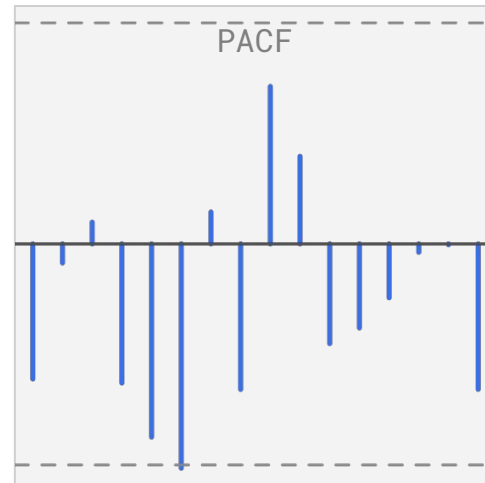
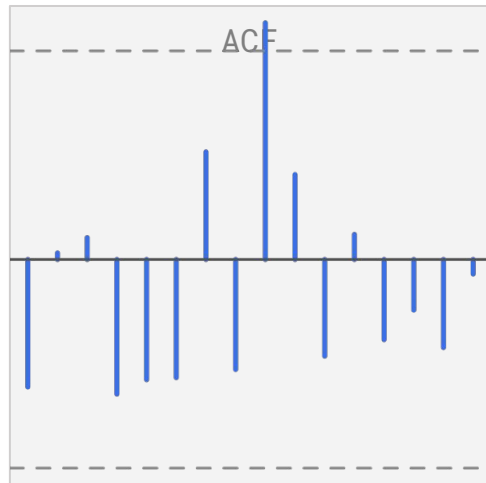


✓ Normality



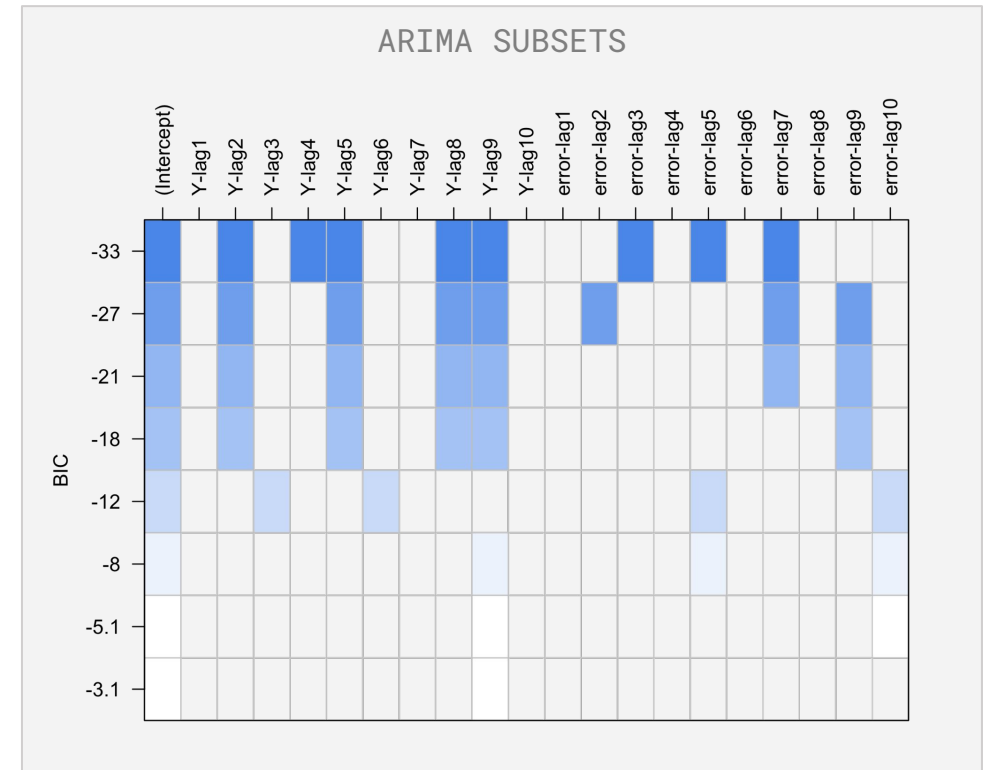
✓ Independence

# diff log Price, Candidate Models



**ARIMA(0,0,0)**

BIC: 26.5878



**ARIMA(9,0,7)**

BIC: 52.24535

## diff log Price, Overfitting

ARIMA(0, 0, 0)

<b>p + 1</b>	ARIMA( <b>1</b> , 0, 0):	-0.20 - 0.32	AR(1) LAG	-0.20 + 0.32	✓ Insignificant
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<b>q + 1</b>	ARIMA(0, 0, <b>1</b> ):	-0.21 - 0.33	MA(1) LAG	-0.21 + 0.33	✓ Insignificant
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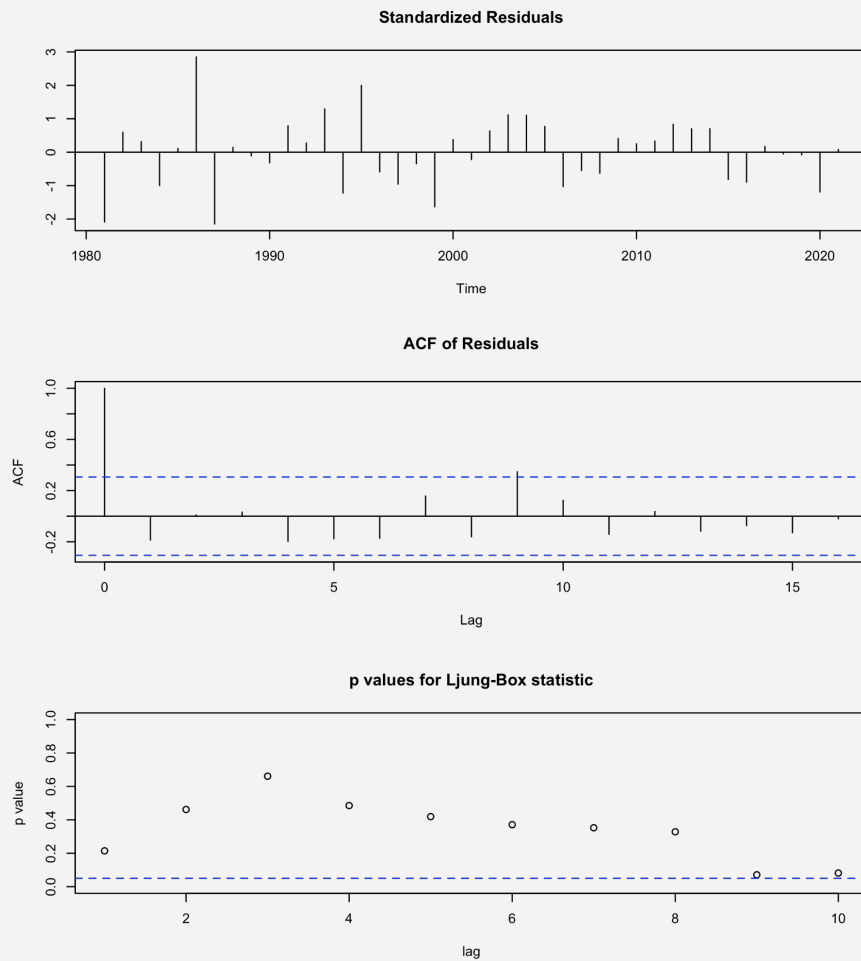
ARIMA(9, 0, 7)

<b>p + 1</b>	ARIMA( <b>10</b> , 0, 7):	-0.17 - 0.60	AR(10) LAG	-0.17 + 0.60	✓ Insignificant
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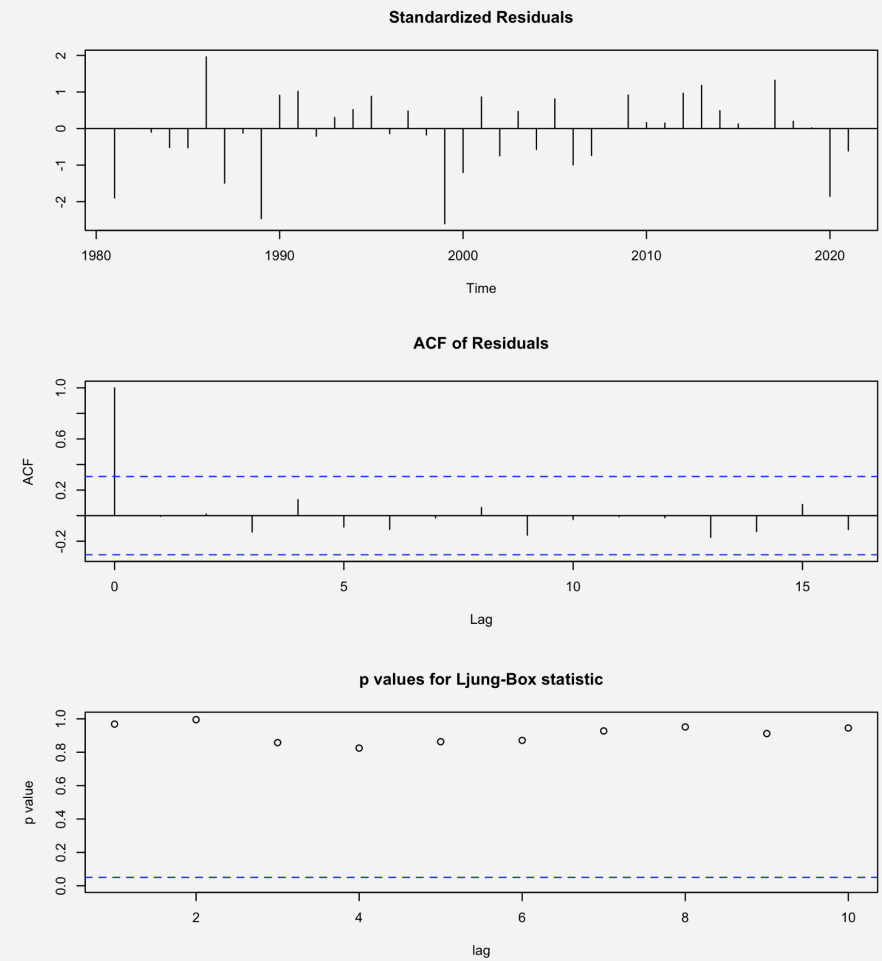
<b>q + 1</b>	ARIMA(9, 0, <b>8</b> ):	-0.95 - 1.29	MA(8) LAG	-0.95 + 1.29	✓ Insignificant
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# diff log Price, ARIMA Residual Analysis

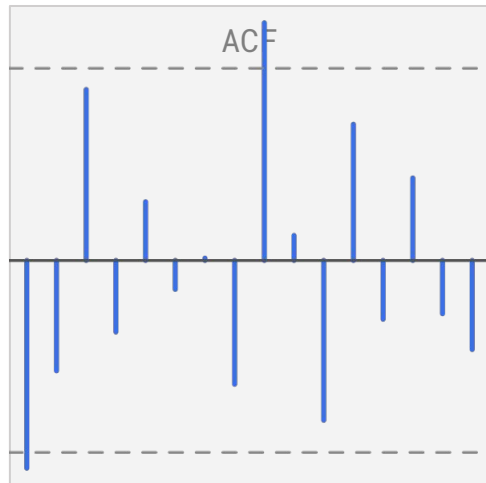
✓ ARIMA(0, 0, 0)



✓ ARIMA(9, 0, 7)

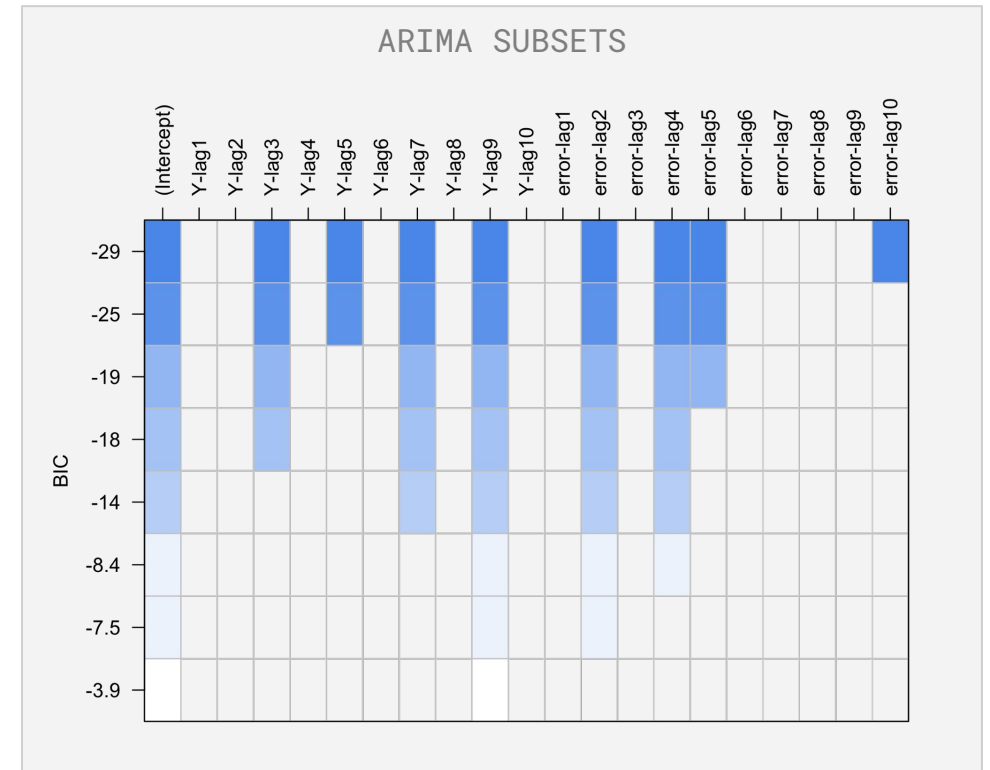
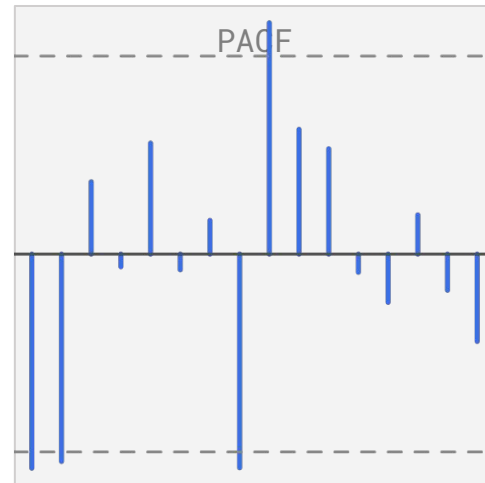


# diff Production, Candidate Models



**ARIMA(0,0,1)**

BIC: 559.2685



**ARIMA(9,0,0)**

BIC: 569.1044



## diff Production, Overfitting

ARIMA(0, 0, 1)

p + 1 ARIMA(1, 0, 1): -0.02 - 0.59 AR(1) LAG -0.02 + 0.59 ✓ Insignificant

q + 1 ARIMA(0, 0, 2): +0.04 - 0.49 MA(2) LAG +0.04 + 0.49 ✓ Insignificant

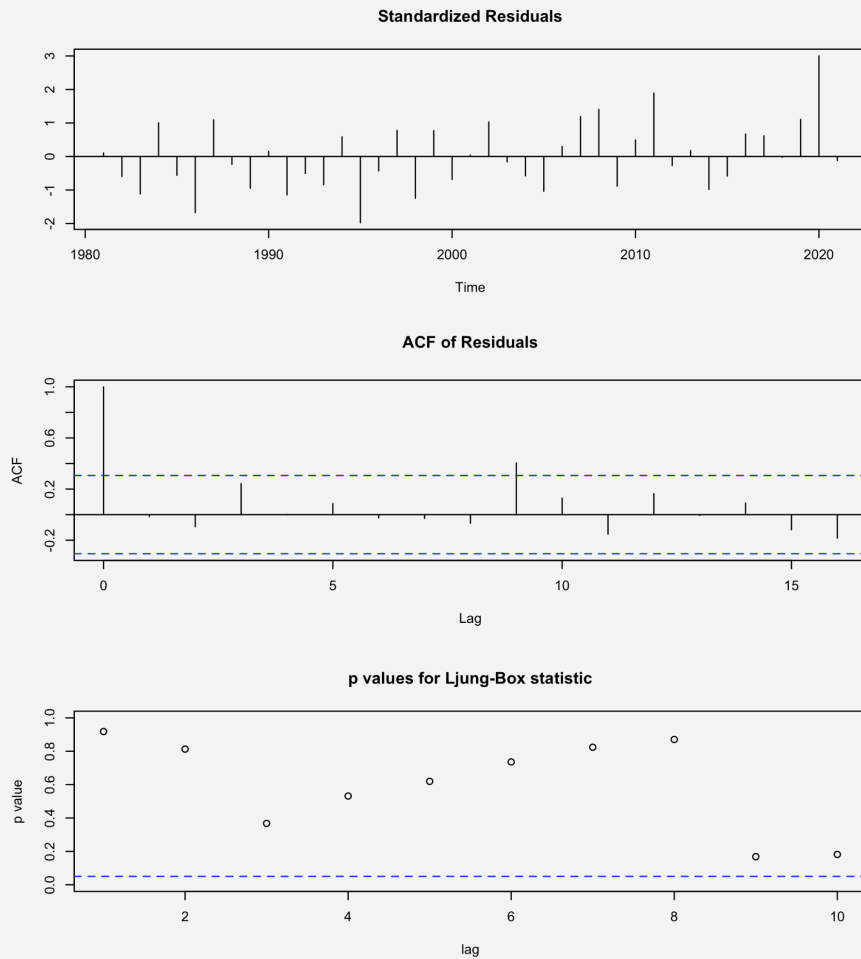
ARIMA(9, 0, 0)

p + 1 ARIMA(10, 0, 0): +0.29 - 0.33 AR(10) LAG +0.29 + 0.33 ✓ Insignificant

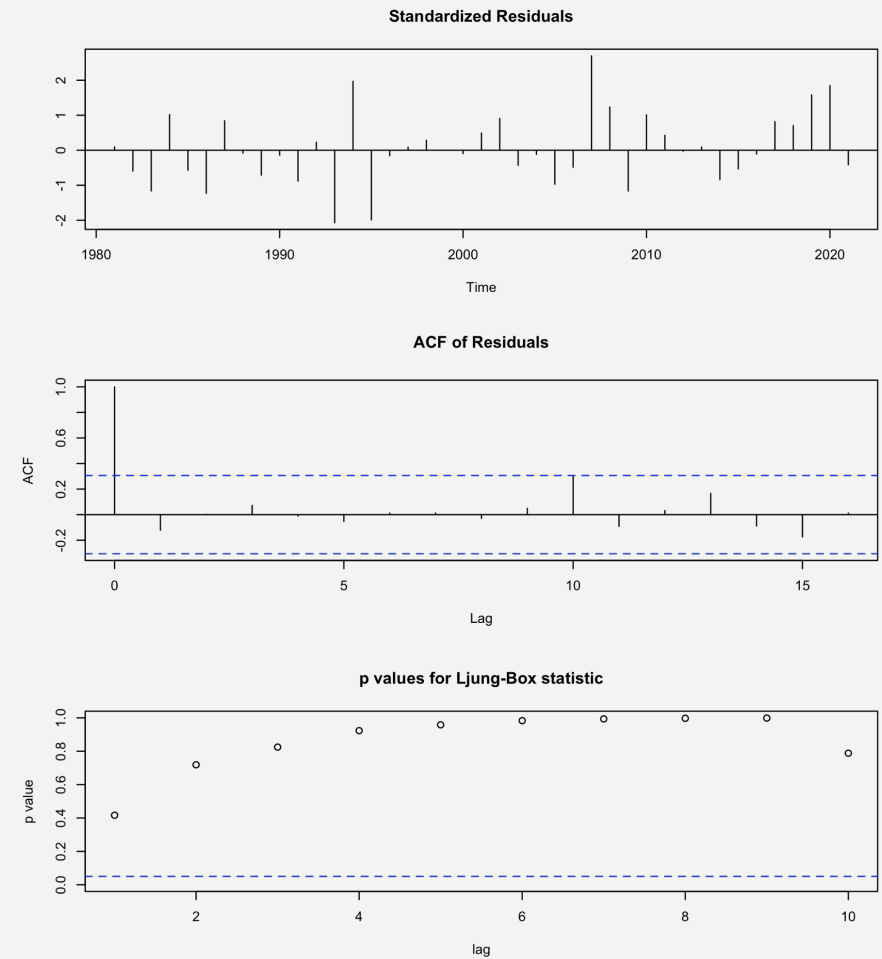
q + 1 ARIMA(9, 0, 1): -0.52 - 0.37 MA(1) LAG -0.52 + 0.37 ✗ Significant

# diff Production, ARIMA Residual Analysis

✓ **ARIMA(0, 0, 1)**



✓ **ARIMA(9, 0, 0)**



# And now, the weather.

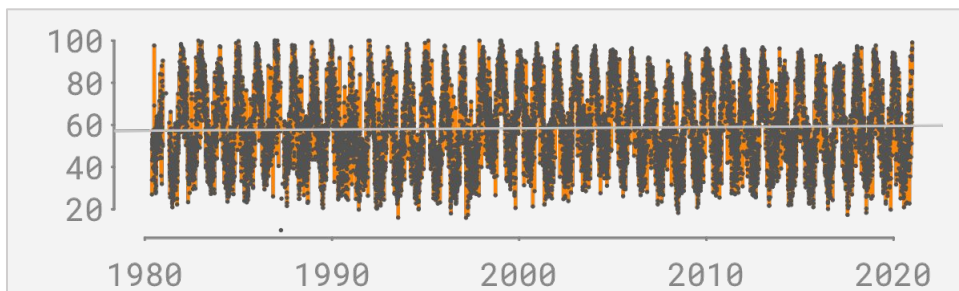
**Humidity**

**Temperature**

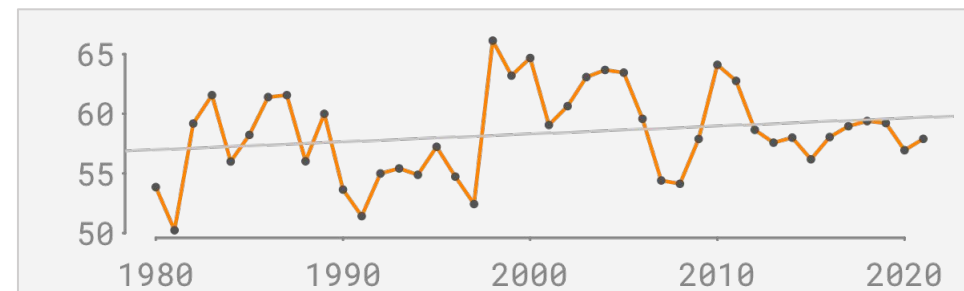
**Windspeed**

# Weather

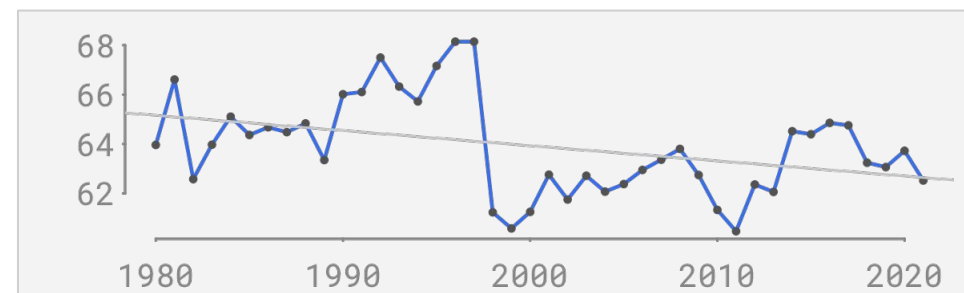
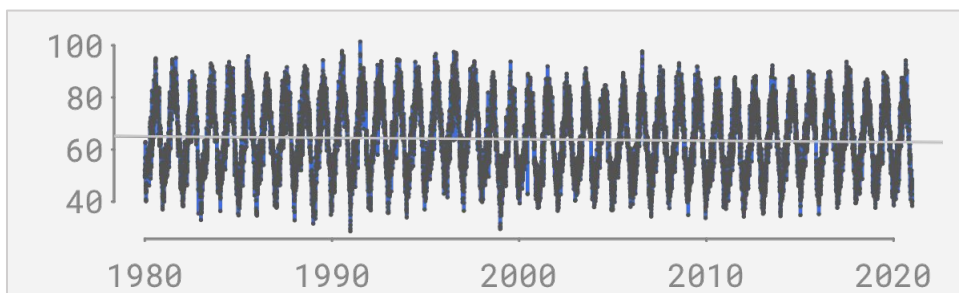
H



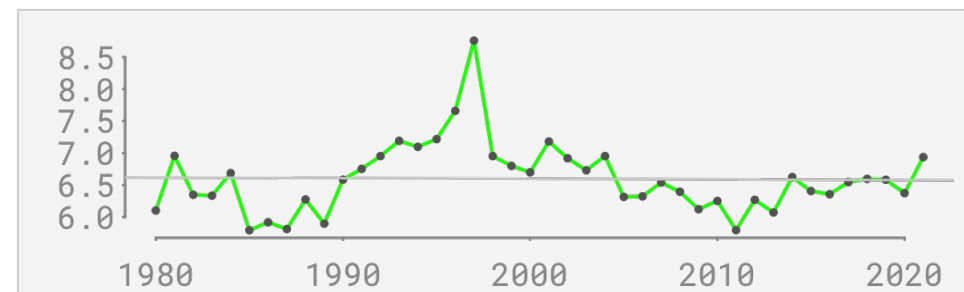
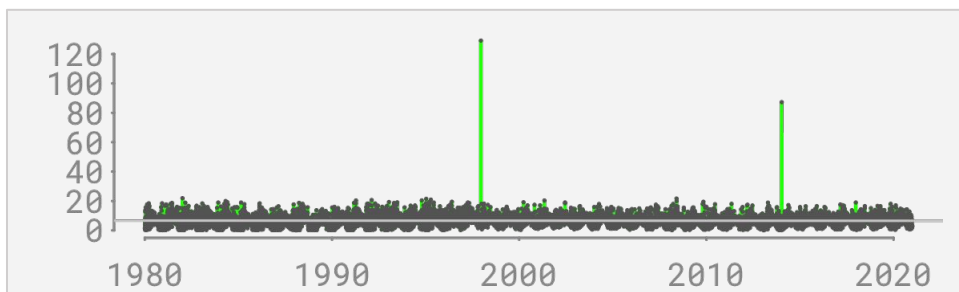
## yearly avg Weather\*



T



W



# diff log **Price, Weather Covariates**

				BIC
ARIMA(0,0,0)	H	T	W	29.48383
	H	T		<b>25.96474</b>
	H		W	27.95859
		T	W	33.75498
	H			<b>25.74740</b>
		T		30.22771
			W	30.05483
				<b>26.58779</b>

				BIC
ARIMA(9,0,7)	H	T	W	55.94549
	H	T		58.31152
	H		W	54.72239
		T	W	58.15431
	H			55.92905
		T		59.17534
			W	57.97431
				<b>52.24535</b>

# diff Production, Weather Covariates

				BIC
ARIMA(0,0,1)	H	T	W	560.3088
	H	T		<b>556.7073</b>
	H		W	559.8764
		T	W	566.5588
	H			<b>557.6343</b>
		T		562.9416
			W	562.8474
				<b>559.2685</b>

				BIC
ARIMA(9,0,0)	H	T	W	573.2787
	H	T		570.4709
	H		W	570.3970
		T	W	573.1860
	H			<b>566.9554</b>
		T		570.2323
			W	569.7328
				<b>569.1044</b>

# C.I.'s for Weather Condition Coefficients

diff log **Price**

ARIMA(0, 0, 0)	H		$0.0251 \pm 0.0115$	✓ Significant
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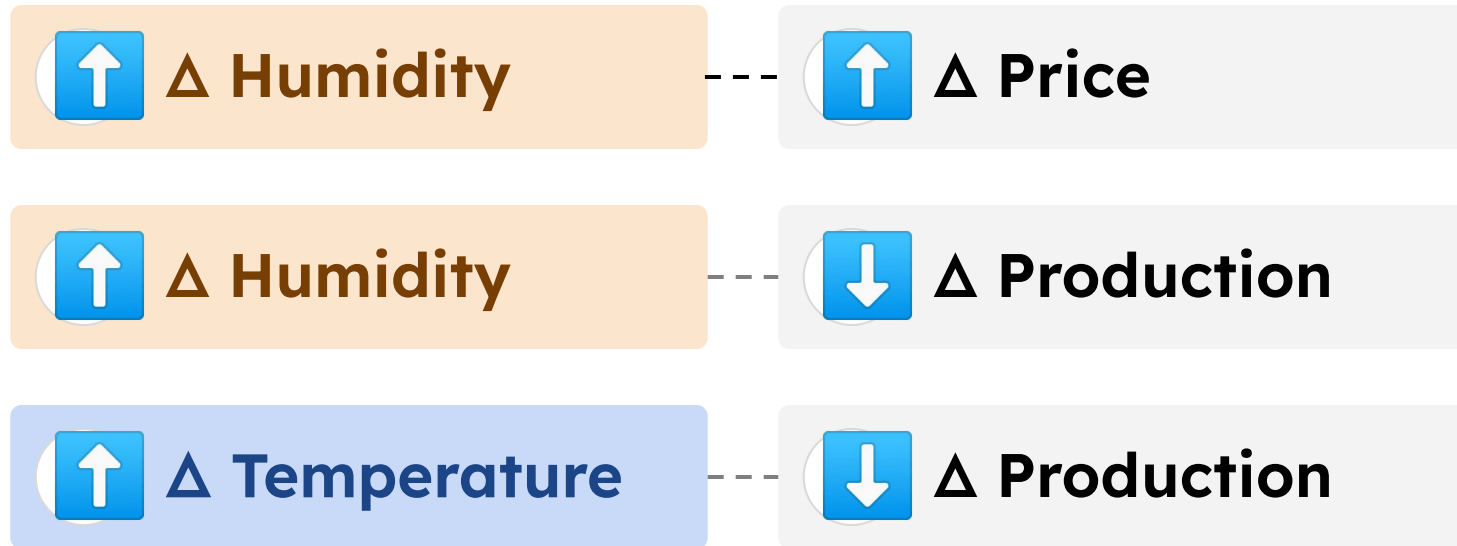
diff **Production**

ARIMA(0, 0, 1)	H	T	$-34.659 \pm 10.335$	$-53.455 \pm 24.019$	✓ Significant
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# Conclusions

- Weather regressors reduce model complexity & improve BIC



- Potential* feature selection (humidity, humidity and temperature)

