# Can ARIMA predict Currency Exchange Rates?

Euro to US Dollar

by Maren Beckman

## Contents

- Inspiration
- Data
- Time Series Analysis
- Parameters
- Models
- Measuring Success
- Conclusion
- Appendix

# Inspiration

"4 Ways to Forecast Exchange Rates" by Joseph Nguyen from investopedia.com

- No method has proven more successful
- Purchasing Power Parity the economist's choice, compares value of goods (McDonald's Big Mac)
- Relative Economic Strength supplements other methods with interest rate data
- Econometrics incorporated various data estimated to impact currency value

# Inspiration

continued

- Time series uses only historical values to make predictions
- Most popular time series model is ARIMA
- This investigation focuses on ARIMA

## What is ARIMA model?

#### The ARIMA model takes three parameters:

- "AR" for auto-regressive or how the data relates to its prior values
- "I" for integrated or the level of differencing required to make the data stationary
- "MA" for moving average or how the data trends

# **Data**Source

- Dataset collected from US Federal Reserve Exchange Rates at <a href="https://www.federalreserve.gov/data.htm">https://www.federalreserve.gov/data.htm</a>
- Euros to US Dollars
- Euro introduced in January 2002
- Data through August 2018

## Exploration

	count	mean	std	min	25%	50%	75%	max
Rate	4189	1.25222	0.141234	0.8594	1.1534	1.2684	1.3494	1.601



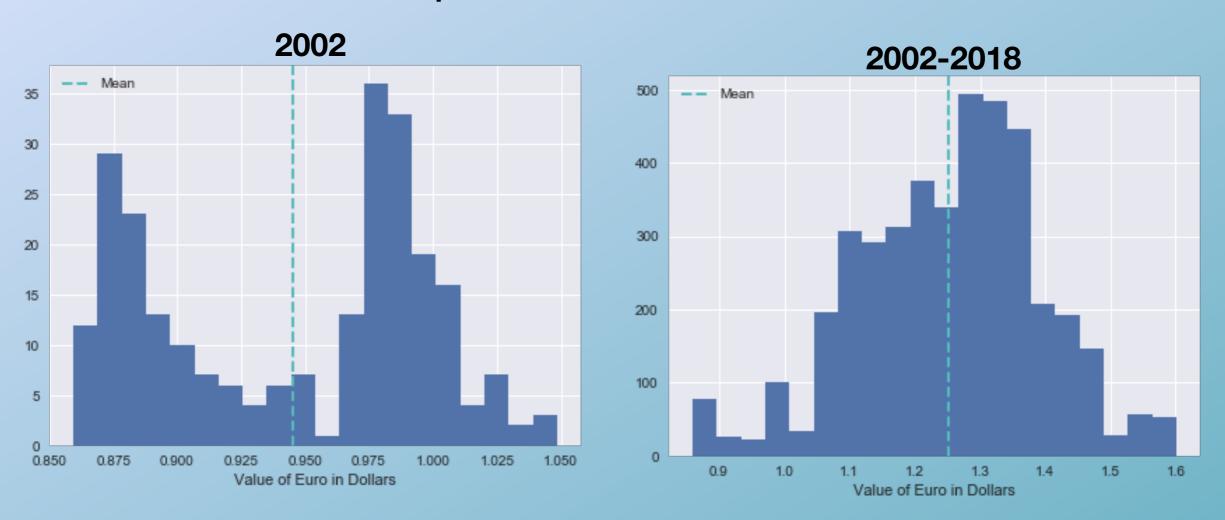
#### **Exploration Continued**



#### Euro value burn-in?

The lowest values are in the first year as the Euro gains confidence against US currency.

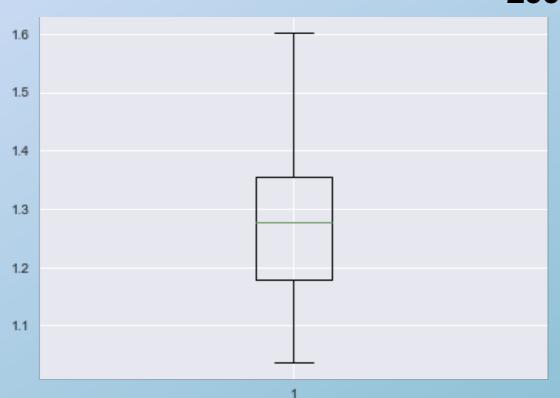
#### **Exploration Continued**

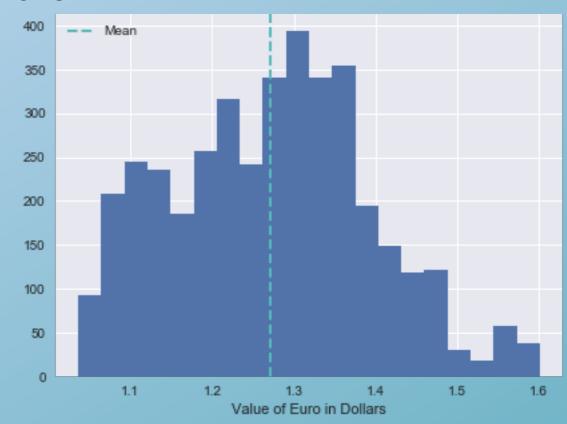


- 2002 values are inconsistent with the rest of the data
- We'll drop the first year of data to get a more established value of the exchange rate

#### **Exploration Continued**

#### 2003-2018





- Mean of 1.27
- Range from 1.04 to 1.60
- No outliers, we have normal distribution

# Time Series Analysis

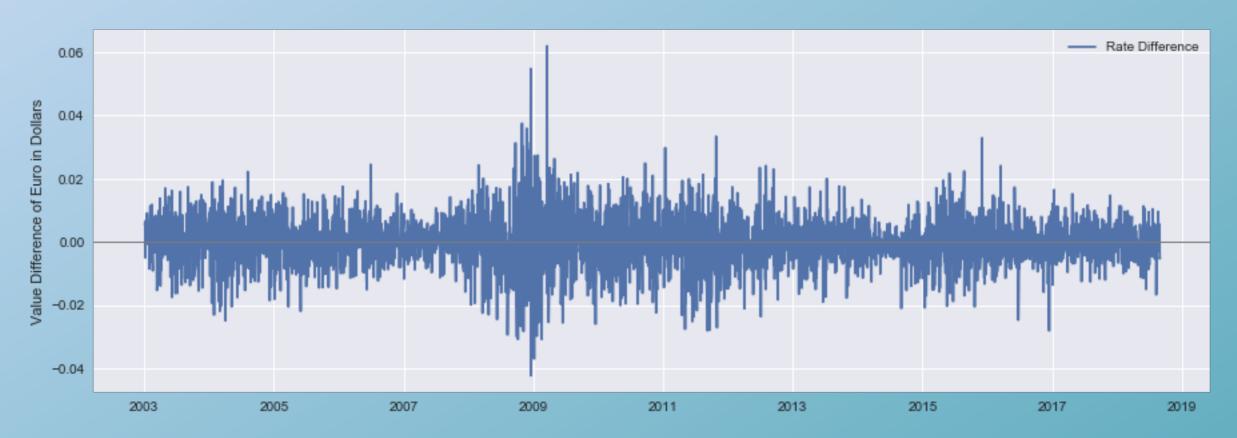


- Final data
- Euro is more valuable than Dollar
- Euro ranges from \$1.04 to \$1.60 (in 2008)

# Analysis

#### Stationarity

- ARIMA needs stationary data
- Take the difference of each data point from the prior value
- Visualization of the results of one degree of differencing



• Two degree of differencing was also done

# Stationarity Dickey Fuller Test

	Statistic	p-Value
Original Data	-2.403107	0.140861
First Difference	-62.348214	0.00000
Second Difference	-18.877696	0.00000

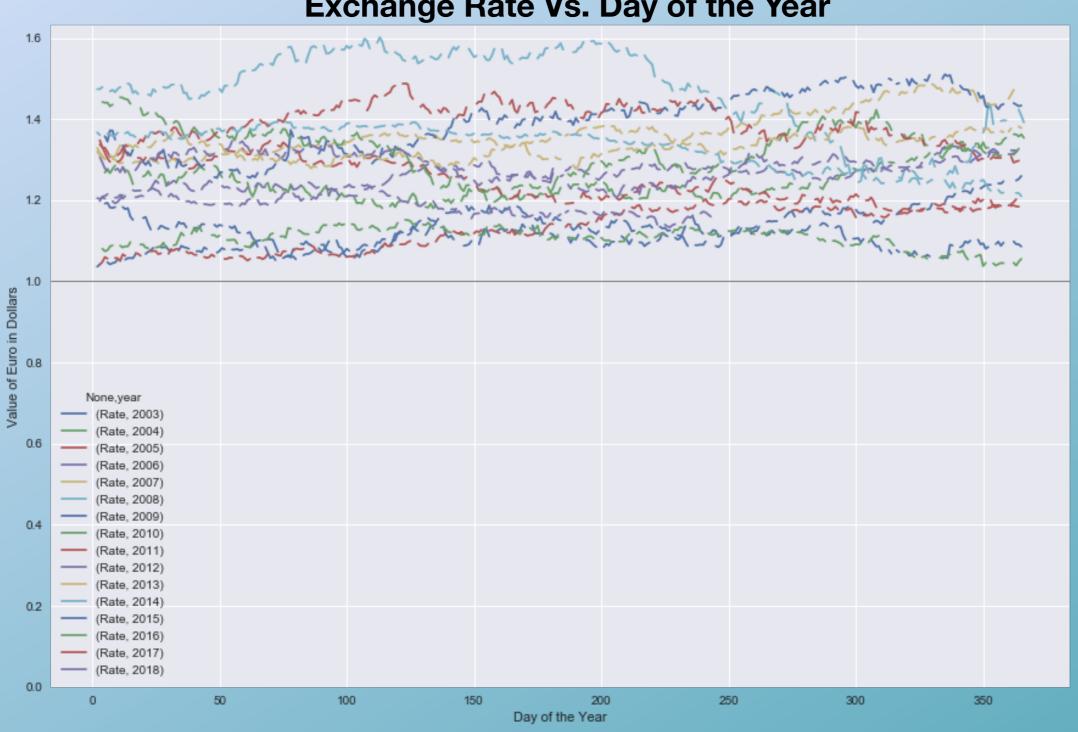
- Negative statistic indicates strength of stationarity
- Small p-value indicates significance of statistic
- First difference lowest statistic and smallest p-value

This confirms that the first difference best establishes stationarity.

# Analysis

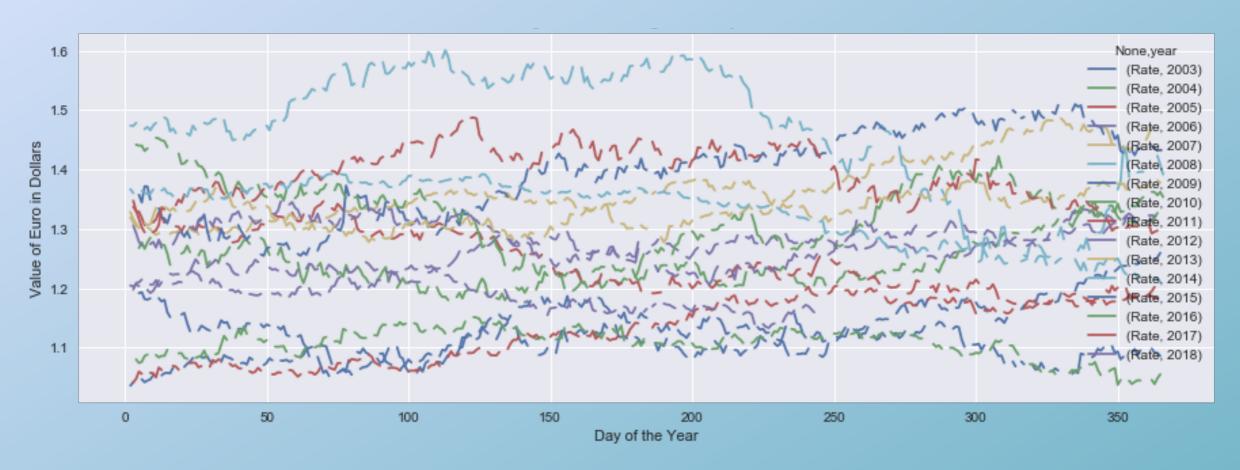
## Seasonality

**Exchange Rate Vs. Day of the Year** 



## Analysis

### Seasonality continued

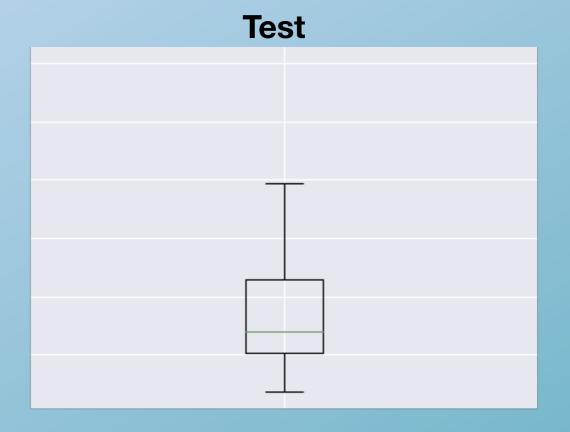


- No visible patterns the records criss cross each other with no repetition - no seasonality
- Non-seasonal ARIMA model will best suit the data.

# **Model Preparation**

70/30 Training/Test

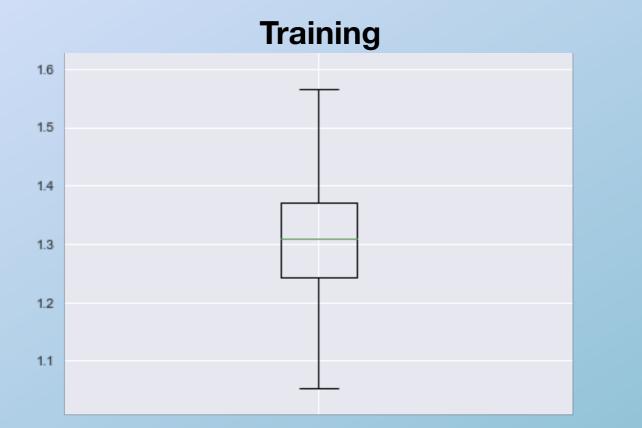


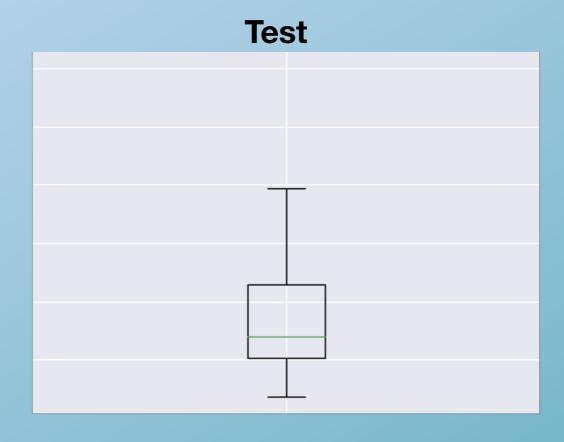


	Count	Start	End	Minimum	Maximum	Median
Training	2756	Jan 2003	Dec 2013	\$1.04	\$1.60	\$1.31
Test	1182	Dec 2013	Aug 2018	\$1.04	\$1.39	\$1.18

# **Model Preparation**

70/30 Training/Test





	Count	Start	End	Minimum	Maximum	Median
Training	2756	Jan 2003	Dec 2013	\$1.04	\$1.60	\$1.31
Test	1182	Dec 2013	Aug 2018	\$1.04	\$1.39	\$1.18

Stronger Dollar value in test set

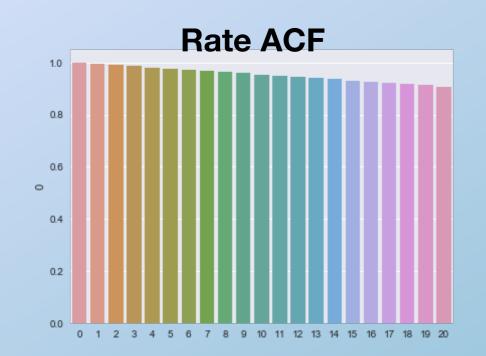
## **ARIMA Model**

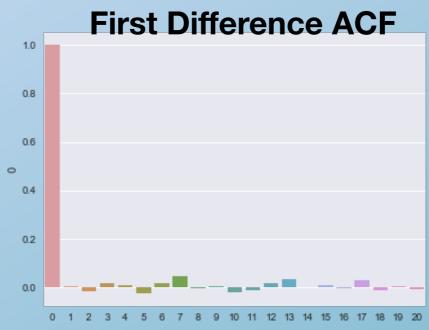
#### **Estimating Parameters**

#### ARIMA parameters to identify:

- Auto-Regressive ('AR') AutoCorrelation Function (ACF) exploring correlations with prior values
- Integrated ('I') Stationarity
- Moving Average ('MA') Partial AutoCorrelation Function (PACF) - exploring isolated correlations between individual values

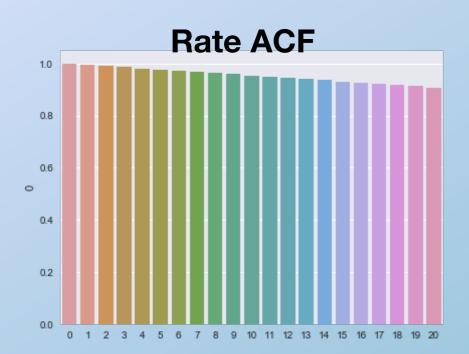
## **ACF** and **PACF**

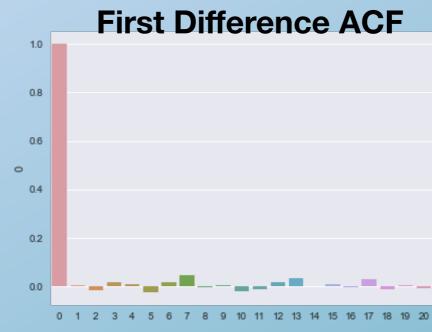




- Differenced data shows no correlation
- No clear AR parameter value

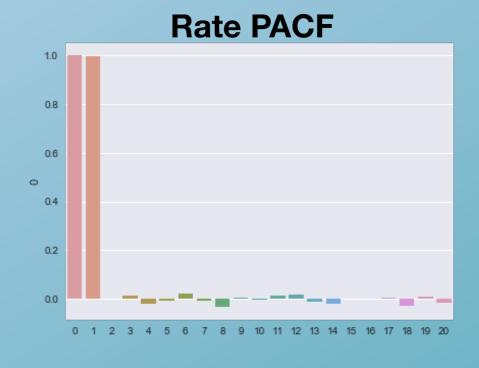
## **ACF** and **PACF**

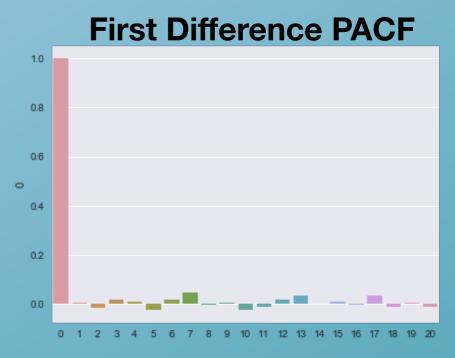




- Differenced data shows no correlation
- No clear AR parameter value

- Differenced data PACF shows no correlation
- No clear MA parameter value





## **ARIMA Model**

#### Testing Parameters - Auto Arima Grid Search

	order		AIC	BIC
(0,	1,	0)	-18441.104	-18429.261
(1,	1,	0)	-18439.164	-18421.401
(0,	1,	1)	-18439.166	-18421.402
3,	1,	2)	-18437.215	-18395.767
(4,	1,	2)	-18435.327	-18387.957
(4,	1,	3)	-18433.663	-18380.372
(1,	2,	1)	-18409.738	-18386.055
(0,	2,	1)	-18410.152	-18392.389
(1,	2,	2)	-18423.269	-18393.664
(2,	2,	2)	-18414.372	-18378.848
(1,	2,	3)	-18404.859	-18369.334
(2,	2,	3)	-18399.697	-18358.251

AIC and BIC metrics balancing complexity with data explanation

Each model tested captures the data with similar accuracy and parsimony.

# **Model Training & Prediction**

Testing model efficacy

Train four models using different parameters:

• (0, 1, 0)

• (3, 1, 2)

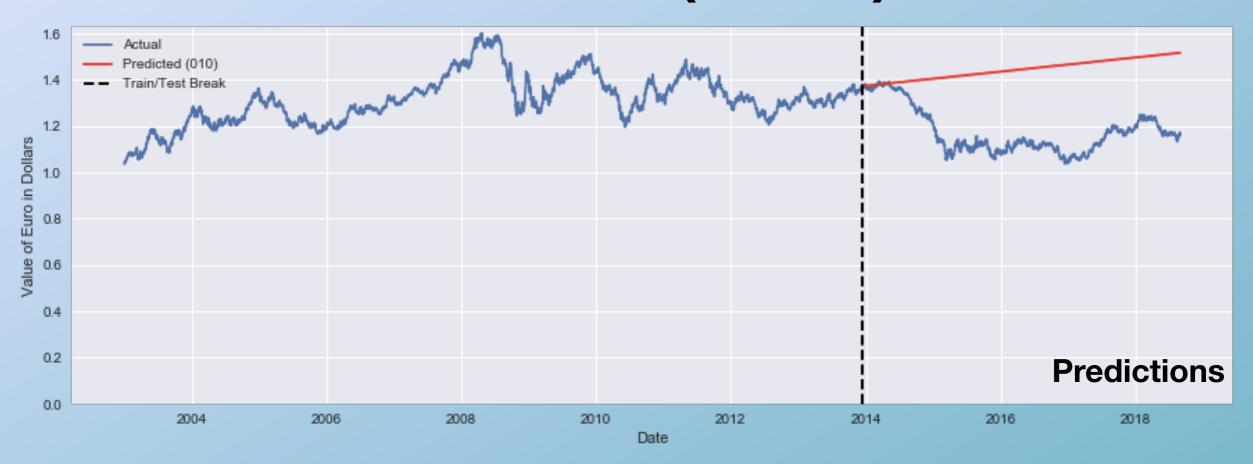
• (0, 1, 1)

• (1, 2, 2)

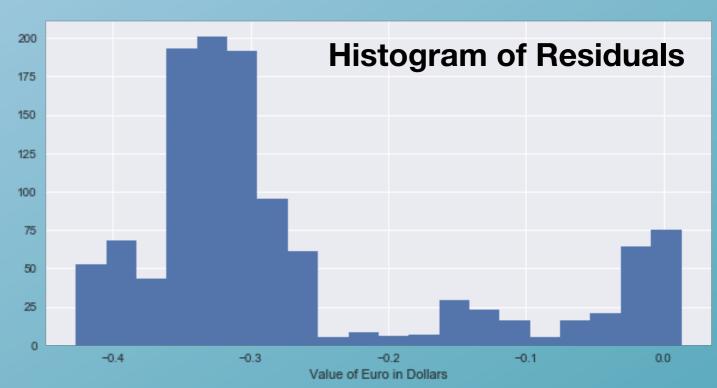
#### **Out of Sample Forecasting**

Predicting values for the test range and comparing those results to the actual values.

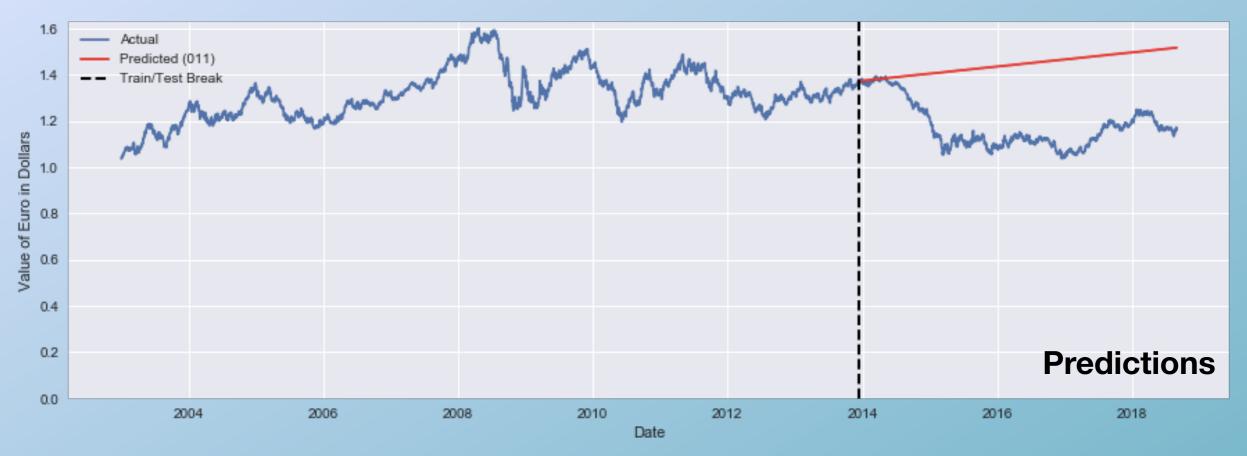
# ARIMA (0,1,0)



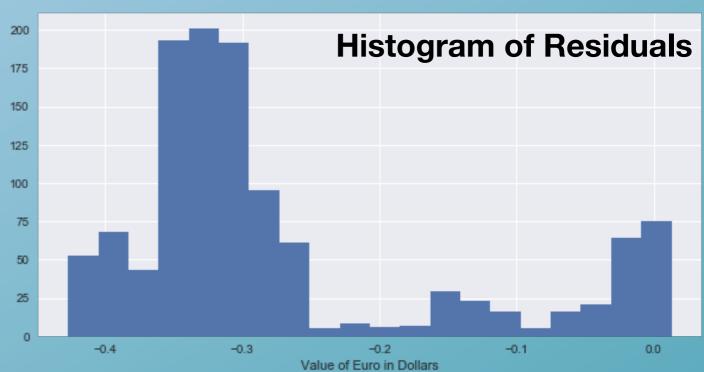
- Increasing straight line
- Predictions are not very accurate
- Residuals are not normally distributed



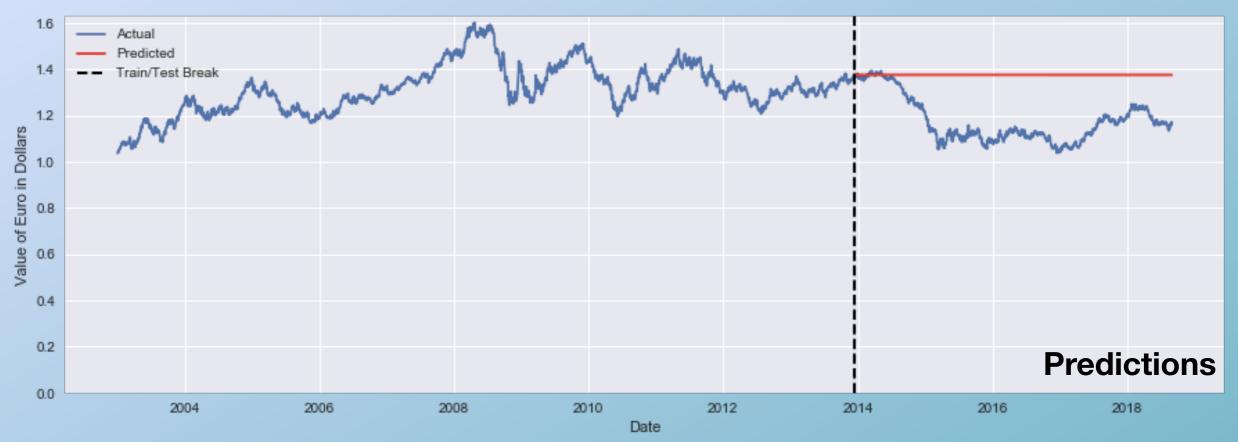
# ARIMA (0,1,1)



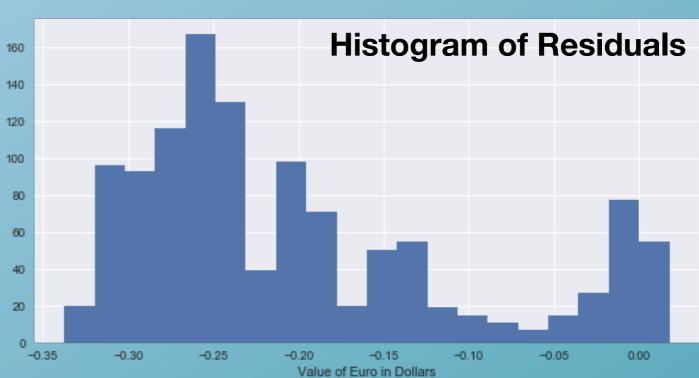
Results nearly identical to (0, 1, 1) model



# ARIMA (1,2,2)



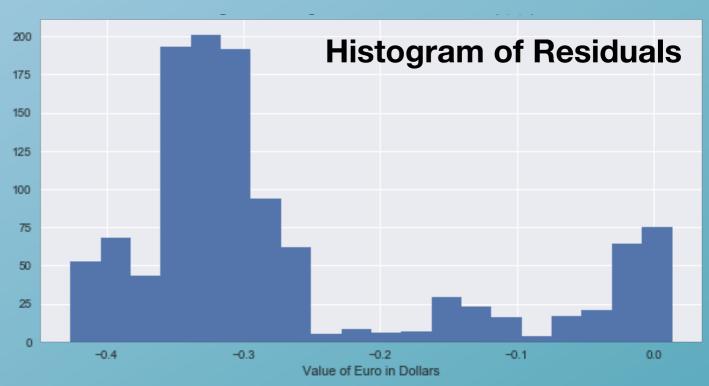
- Flat straight line
- Reduced residual range, but still not normally distributed



# ARIMA (3,1,2)



- Increasing straight line
- Residuals reflect inconsistency with actual values



## **Prediction Success**

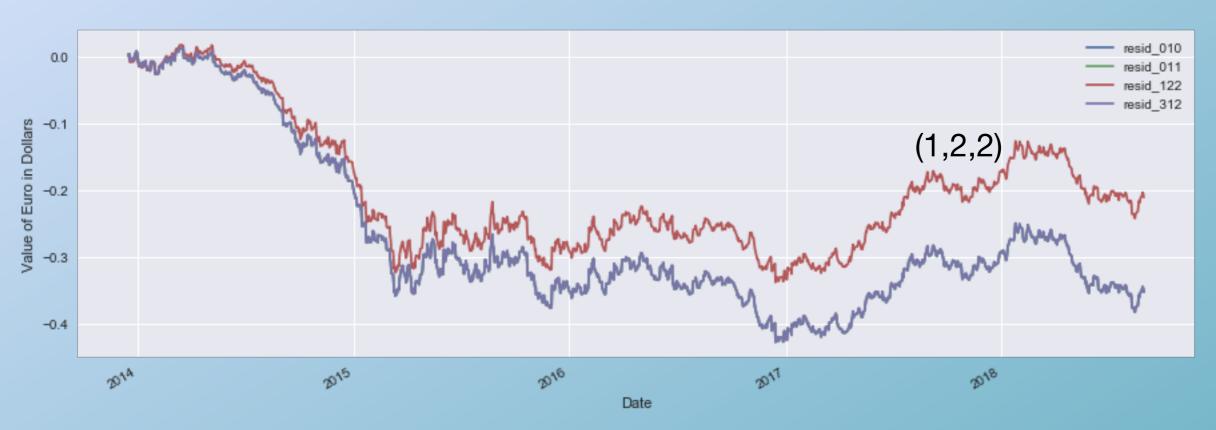
Mean Squared Errors

Model	MSE
(0, 1, 0)	0.087535
(0, 1, 1)	0.087527
(1, 2, 2)	0.049368
(3, 1, 2)	0.087295

- Similar accuracy with models (0,1,0), (0,1,1) and (3,1,2)
- Best accuracy with ARIMA (1,2,2)

## **Prediction Success**

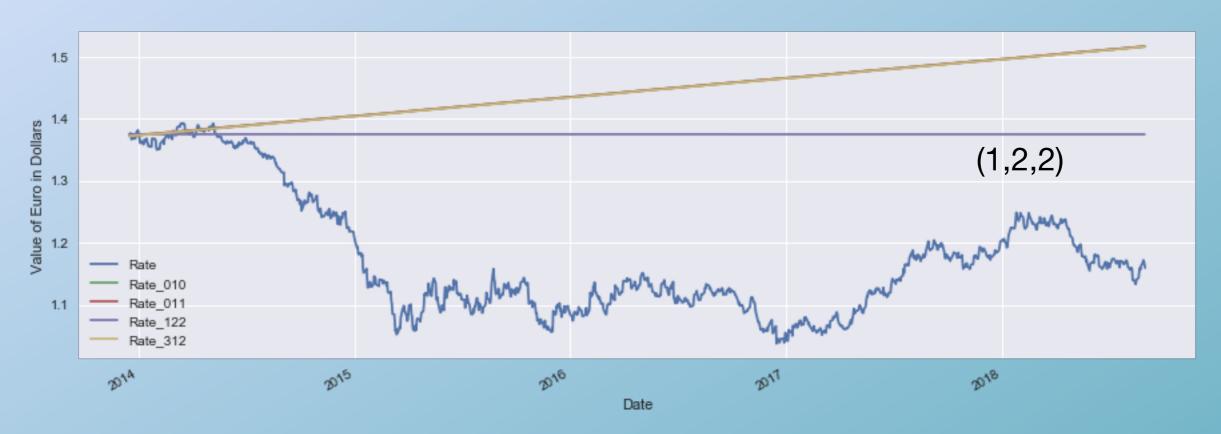
Visualizing Residuals over time



- Residuals are not randomly distributed
- Similar accuracy in the first half-year
- (1, 2, 2) has comparatively decreasing residuals

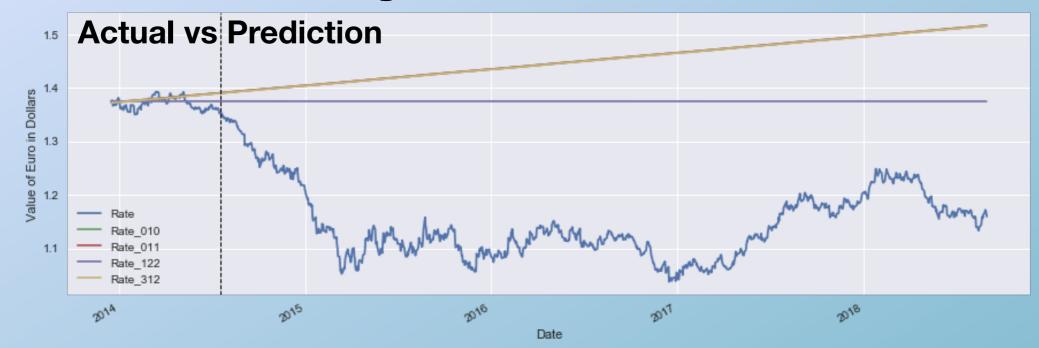
## **Prediction Success**

Visual Comparison of Actual vs Prediction

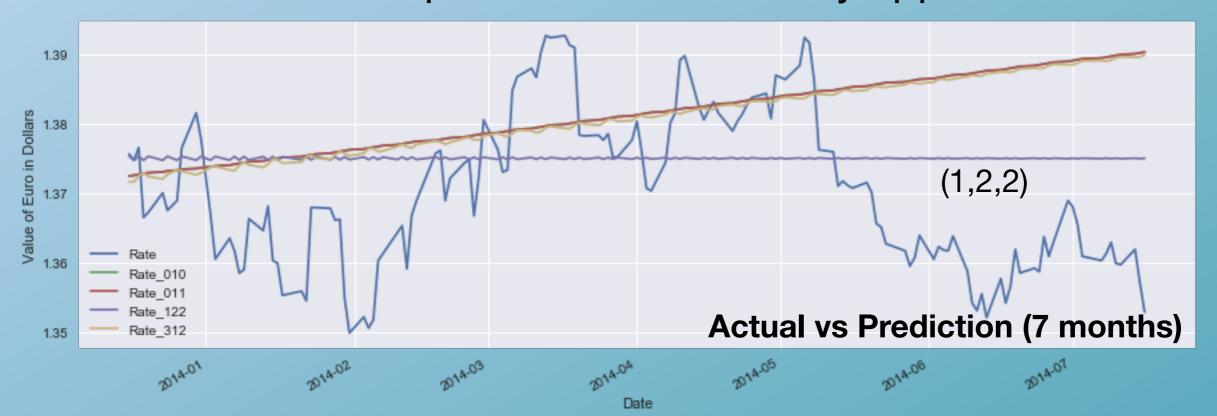


- Three of four models predicted increasing rates
- Actual Euro value had a dramatic decrease in the first year
- Model (1,2,2) appears to be most accurate

# **Early Predictions**



- Euro value drops about seven months into the test data
- Trends in model predictions are already apparent



# **Early Prediction Success**

Model	MSE
010	0.000282
011	0.000282
122	0.000143
312	0.000273



## Conclusion

- ARIMA model applied to exchange rate of Euro to US Dollar
- Dickey Fuller Test, ACF and PACF used to target model parameters
- Four Models selected using Auto ARIMA grid search:
   (1, 2, 2), (3, 1, 2), (0, 1, 0), (0, 1, 1)
- Best model (1, 2, 2) Iowest Mean Squred Error (MSE)
- Observed weak fit residuals were not randomly distributed, histogram of residuals was not normally distributed
- Poor performance over five year test period
- Acceptable accuracy with predictions on first seven months

## **Further Exploration**

#### **Facebook's Prophet**

Additive model gaining in popularity.

#### **Bayesian Structural Time Series Models**

 Greater transparency without differencing, lags and moving averages (suggested by Kim Larsen from Stitch Fix)

#### Linear Regression (Boosted Least Squared Regression)

- Benefit of normal distribution of the data
- Create features and predict future values without additional data

#### **Rolling Forecast**

- Add new data as available and update predictions
- ARIMA likely to show better accuracy in short-term predictions

#### References

4 Ways to Forecast Exchange Rates by Joseph Nguyen https://www.investopedia.com/articles/forex/11/4-ways-to-forecast-exchange-rates.asp

#### Big Mac PP

https://www.investopedia.com/terms/b/bigmacppp.asp

What is an Autoregressive Integrated Moving Average - ARIMA https://www.investopedia.com/terms/a/autoregressive-integrated-moving-average-arima.asp

**US Federal Reserve Exchange Rates** 

https://www.federalreserve.gov/data.htm

Plot each year of a time series on the same x-axis using Pandas by Andrew Tedstone http://atedstone.github.io/pandas-plot-seasons-time-series/

A Gentle Introduction to Autocorrelation and Partial Autocorrelation
https://machinelearningmastery.com/gentle-introduction-autocorrelation-partial-autocorrelation/

A comprehensive Beginner's Guide to Time Series Forecasting by Aarshay Jain https://www.analyticsvidhya.com/blog/2016/02/time-series-forecasting-codes-python/

Sorry ARIMA, but I'm Going Bayesian by Kim Larsen https://www.predictiveanalyticsworld.com/patimes/sorry-arima-but-im-going-bayesian/7565/

Prediction of Foreign Exchange Rate using Regression Techniques http://mtmi.us/rbtr/sept2017/06-Sharma-Hota-Handa-pp29-33.pdf

*Time Series Analysis in Python* by William Koehrsen https://towardsdatascience.com/time-series-analysis-in-python-an-introduction-70d5a5b1d52a