



# Bike Sharing Time Series Analysis

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## *Group Members:*

*Abby(Jingyi) Liu*

*Qianying Diao*

*Qiang Wang*

*Xinrong Chen*

*Jamie(Xiaojie) Pan*



# Agenda

**Introduction  
and Overview**

**Univariate Models**

**Multivariate Models**

**TF Model  
with regressors**

**Vector Model**



Part **1**

# Introduction & Overview



# Introduction

## Background Overview

- **Bike Sharing System** is an innovation of traditional vehicles rental services.
- **Daily counts** of rental bikes from **2011.1.1 to 2012.12.31** in Capital Bikeshare System in Washington D.C..
- Also includes daily weather, humidity, wind speed, precipitation.

## Variable Description

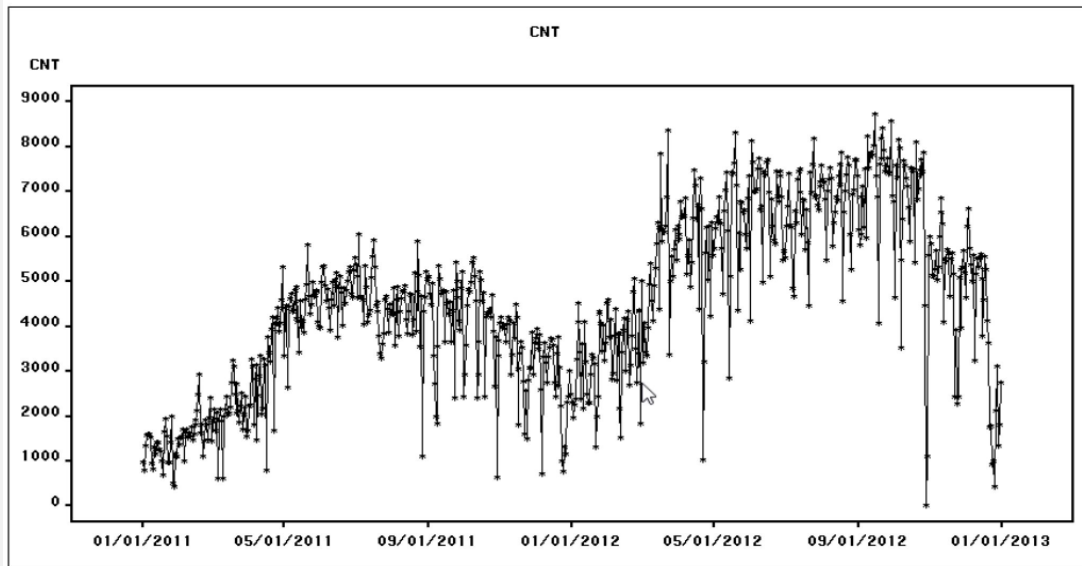
### Dependent Variable

- **cnt**: Count of total rental bikes trips including both casual and registered

### Independent Variables

- **atemp**: Normalized feeling temperature in Celsius.
- **hum**: Normalized humidity.
- **windspeed**: Normalized wind speed.
- **prcp**: historical daily precipitation observations.
- **start\_holiday**: dummies variable, 1 for previous day is working day current day is holiday.
- **end\_holiday**: dummies variable, 1 for previous day is holiday current day is work day.

# Introduction



## Features:

- Nonstationary series
  - Need first difference
- Increasing change in the variability of the series over time
  - a log transformation can be applied
- Clear seasonality

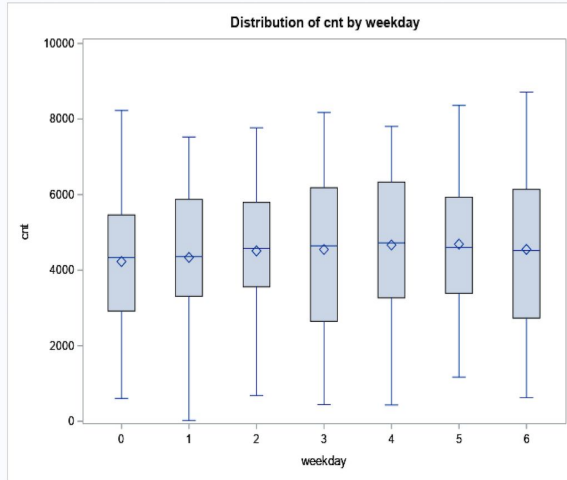


## Part 2

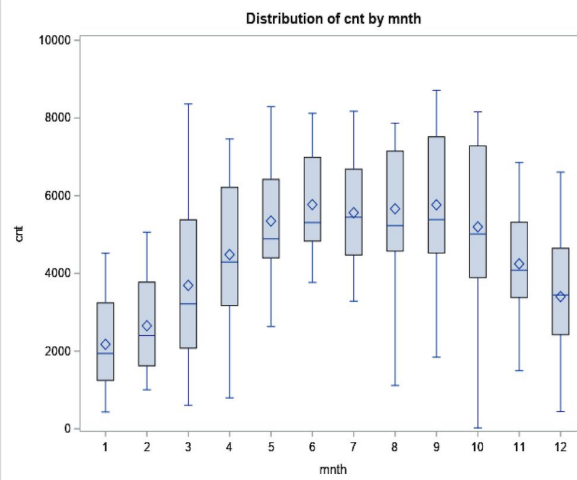
# Univariate Models

# Univariate Models - Seasonal dummies

The SAS System



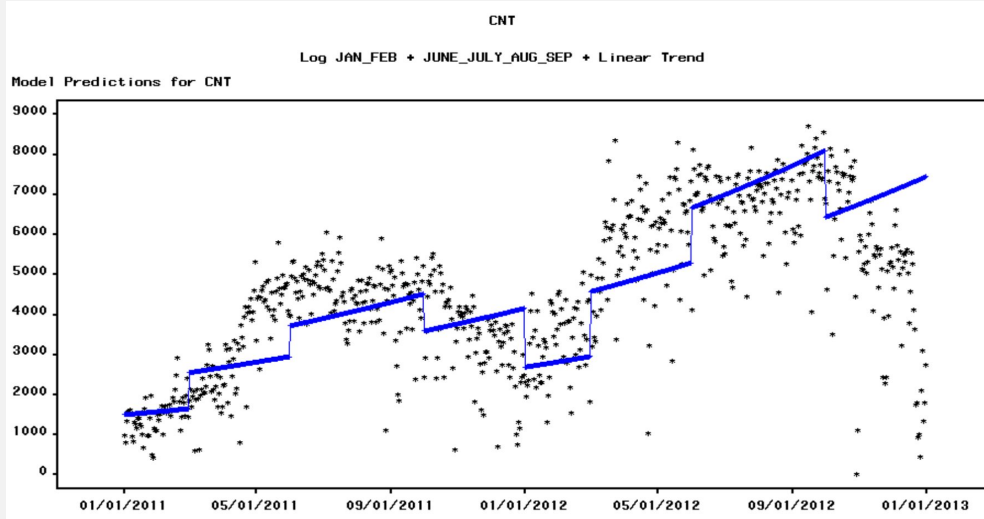
The SAS System



- Create two dummies for peak and bottom months
- 'JAN\_FEB' dummies distinguish Jan. and Feb with other months
- 'JUNE\_JULY\_AUG\_SEP' dummies distinguish June, July, August, and September with other months.

- Weekdays: No significant changes --- No model should be built
- Months: Peak: June to September. Bottom: January and February -- Create dummies

# Univariate Models - Seasonal dummies



## Model Prediction

The model generally grasp most of the ups and downs.

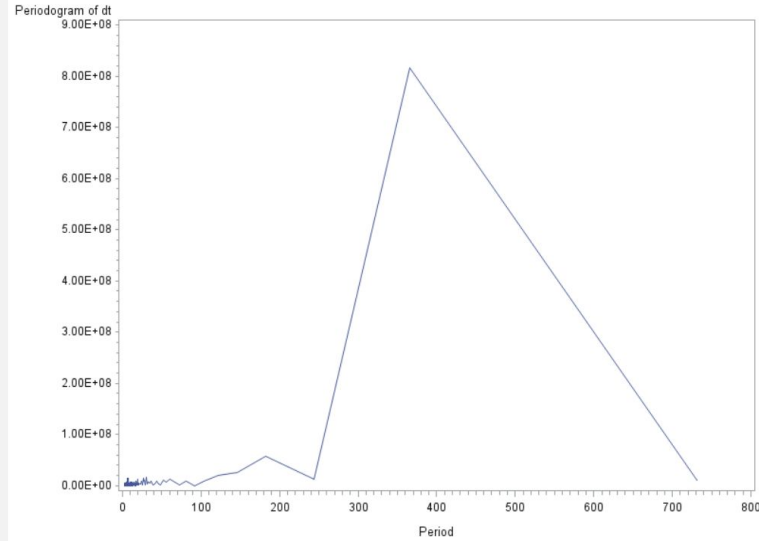
## Parameter Estimate

All parameters are significant

Parameter Estimates				
CNT				
Log JAN_FEB + JUNE_JULY_AUG_SEP + Linear Trend				
Model Parameter	Estimate	Std. Error	T	Prob> T
Intercept	7.75107	0.0311	249.4030	<.0001
JAN_FEB	-0.43567	0.0376	-11.5796	<.0001
JUNE_JULY_AUG_SEP	0.22996	0.0307	7.4814	<.0001
Linear Trend	0.00159	0.000079	20.2492	<.0001
Model Variance (sigma squared)	0.11403	.	.	.



# Univariate Models - Cyclical



	Obs	FREQ	PERIOD	P_01
2	3	0.01719	365.5	816186123
3	5	0.03438	182.75	58007683.1
4	6	0.04298	146.2	25420661.9
5	7	0.05157	121.833	20569437.7
6	25	0.20629	30.458	15529566
7	110	0.93689	6.706	15052975.3
8	29	0.24067	26.107	14403743.3
9	106	0.90251	6.962	14288650.9
10	13	0.10314	60.917	13474732.8
11	39	0.32662	19.237	12266775.6

Periodogram

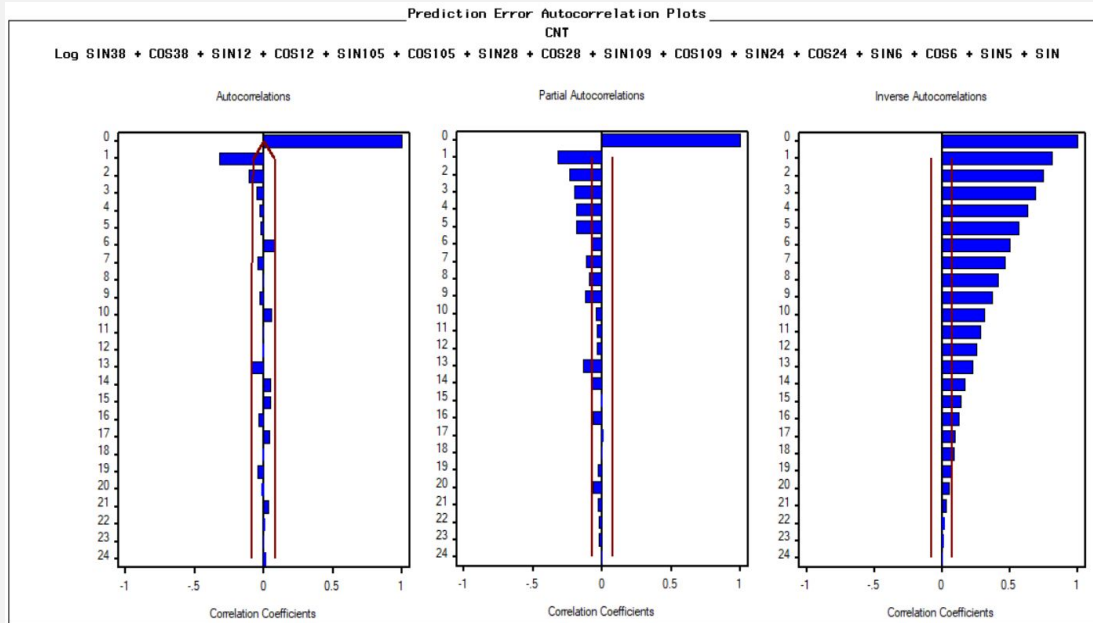


Highest 10 amplitudes



- Non-stationary
- Not White Noise
- **RMSE = 2615.7**

# Univariate Models - Cyclical + First Difference



Take First Difference



- Stationary
- Not WN
- **RMSE = 1310.7**



- ACF decays quickly
- PACF decays quickly
- ARMA(1,1) for Error Model

# Univariate Models - Cyclical + First Difference + Error Model

## Statistic of Fit

Statistic of Fit	Value
Mean Square Error	1556499.9
Root Mean Square Error	1247.6
Mean Absolute Percent Error	245.70072
Mean Absolute Error	992.69333
R-Square	0.595



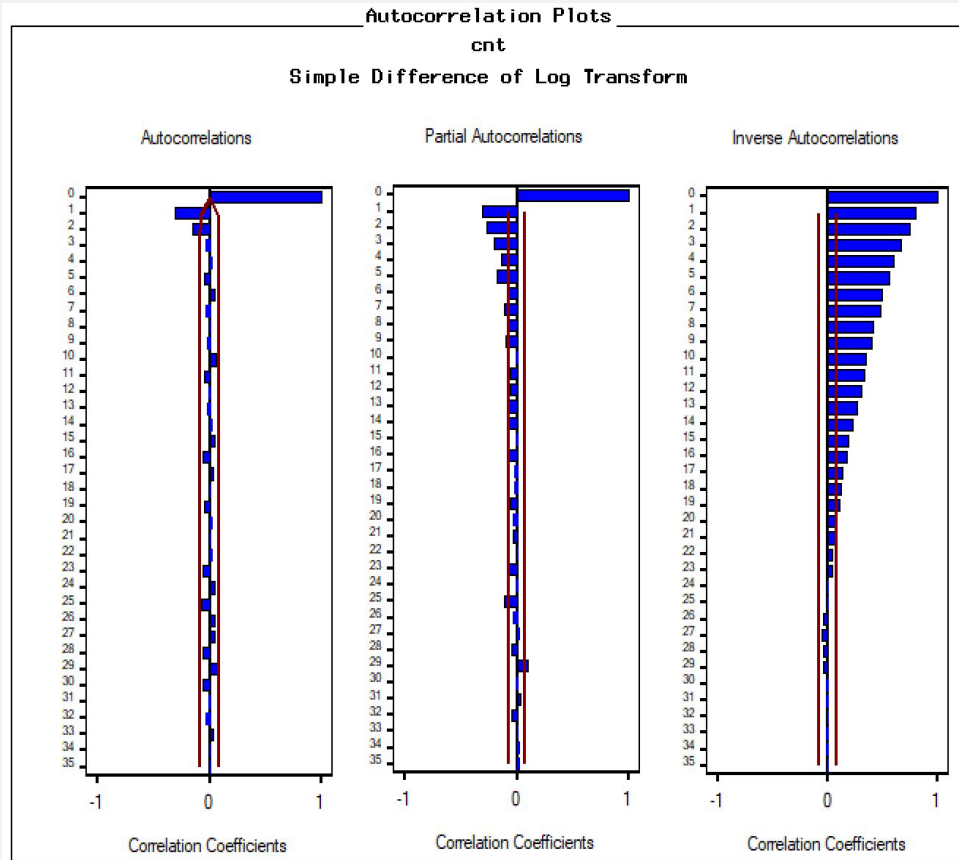
- **Lowest RMSE = 1247.6**
- **Low MAPE = 245.7**
- **Pass WN test**
- **Significant p-values**
  - COS 2 - 1 year
  - COS 4 - 1/2 year
  - COS 109 - 1 week

## Parameter Estimate

Model Parameter	Estimate	Std. Error	T	Prob> T
Intercept	0.00188	0.000512	3.6719	0.0004
Moving Average, Lag 1	0.96732	0.0113	85.3323	<.0001
Autoregressive, Lag 1	0.24225	0.0415	5.8306	<.0001
SIN38	0.03204	0.0194	1.6549	0.1020
COS38	-0.00183	0.0194	-0.0946	0.9249
SIN12	0.00992	0.0208	0.4777	0.6342
COS12	-0.02415	0.0208	-1.1625	0.2486
SIN105	0.02663	0.0172	1.5482	0.1257
COS105	-0.00443	0.0172	-0.2575	0.7975
SIN28	0.01024	0.0198	0.5179	0.6060
COS28	0.02786	0.0197	1.4126	0.1618
SIN109	0.03947	0.0171	2.3142	0.0233
COS109	0.02318	0.0171	1.3597	0.1779
SIN24	0.03396	0.0199	1.7088	0.0915
COS24	-0.02746	0.0199	-1.3816	0.1711
SIN6	0.03006	0.0251	1.1974	0.2348
COS6	-0.01871	0.0251	-0.7447	0.4587
SIN5	0.02608	0.0277	0.9434	0.3484
SIN4	-0.02891	0.0321	-0.9011	0.3703
COS5	0.05556	0.0270	2.0605	0.0427
COS4	-0.08702	0.0290	-2.9966	0.0037
SIN2	0.02967	0.0480	0.6177	0.5386
COS2	-0.40753	0.0483	-8.4422	<.0001
Model Variance (sigma squared)	0.07276	.	.	.

**ARMA(1,1) fits well !**

# Univariate Models - ARIMA



Original series (CNT) is nonstationary



Take first difference



CNT(1) is stationary

## Univariate Models - ARIMA



### ARIMA(1,1,1)

Statistic of Fit	Value
Mean Square Error	1736875.0
Root Mean Square Error	1317.9
Mean Absolute Percent Error	302.29158
Mean Absolute Error	940.46107
R-Square	0.548

### ARIMA(0,1,2)

Statistic of Fit	Value
Mean Square Error	1756347.5
Root Mean Square Error	1325.3
Mean Absolute Percent Error	302.30539
Mean Absolute Error	953.34912
R-Square	0.543

- Out-out samples - 100 obs
- Log Transformation
- Without intercept
- No obvious seasonality

**ARIMA(1,1,1) is better!**

## Univariate Models - Comparison

Model	Model Variance	Root Mean Square Error	Mean Absolute Percent Error
Seasonal Dummy	0.11403	2651.9	388.38455
Cyclical + ARIMA(1,1,1)	0.07276	1247.6	245.70072
ARIMA(0,1,2)	0.07633	1325.3	302.30539
ARIMA(1,1,1)	0.07666	1317.9	302.29158

\*All models here take the log-transformation.

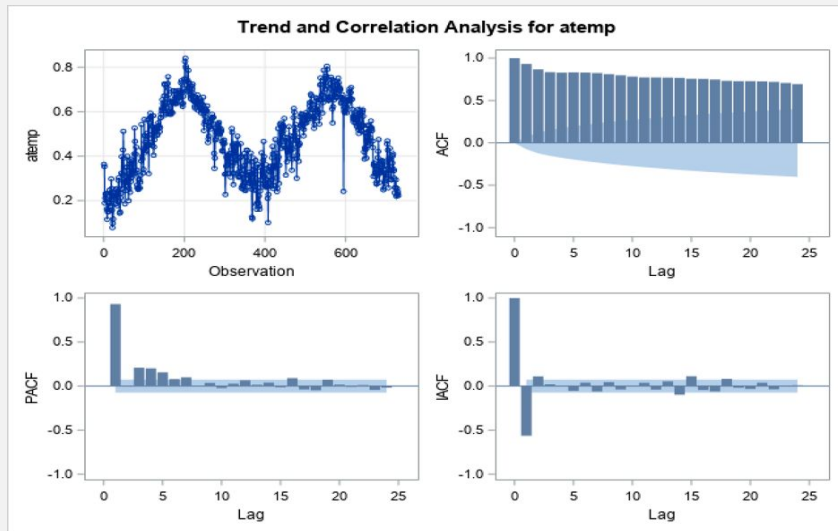


# Part 3

## Multivariate Models

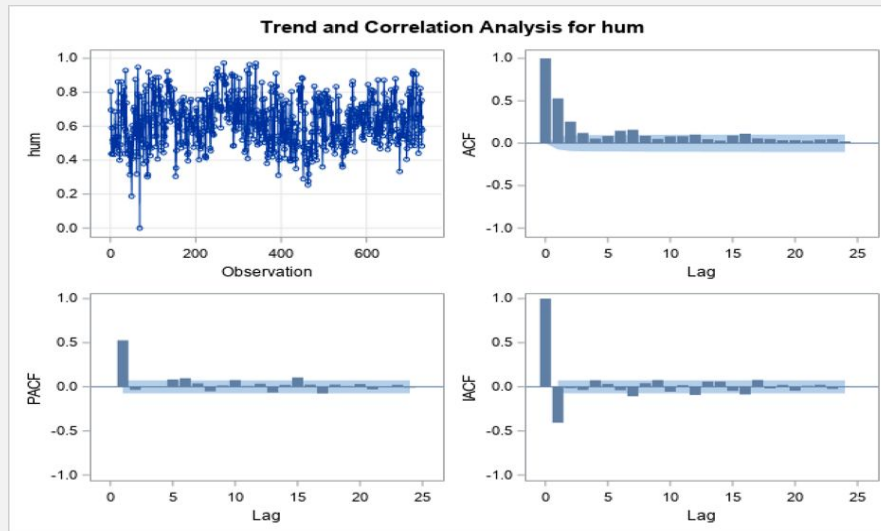
# Multivariate Models - Preview Inputs

## Average Temperature



Nonstationary

## Humidity

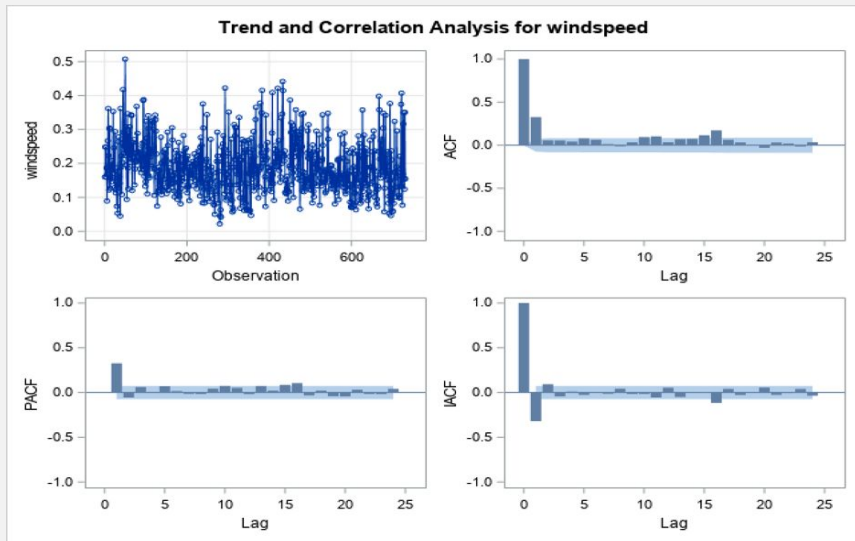


Stationary, NOT White Noise



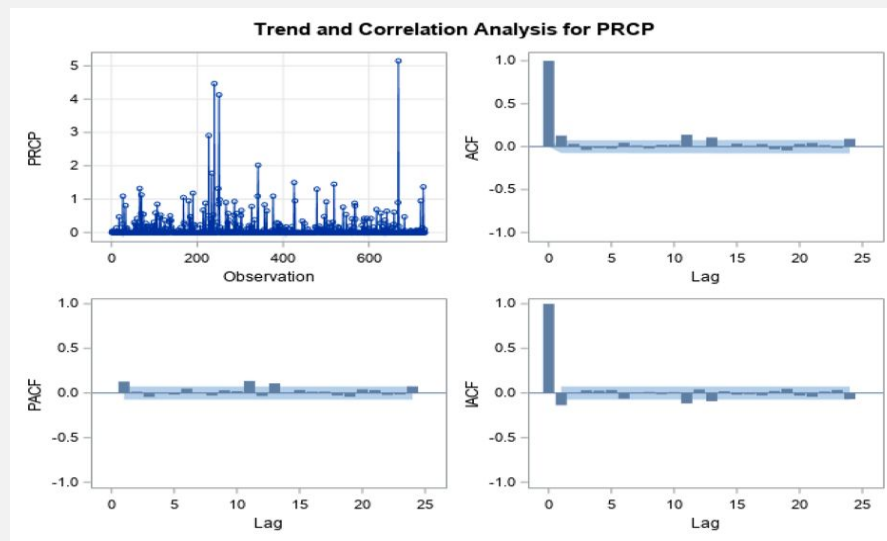
# Multivariate Models - Preview Inputs

## Windspeed



Stationary, NOT White Noise

## Precipitation



Stationary, NOT White Noise

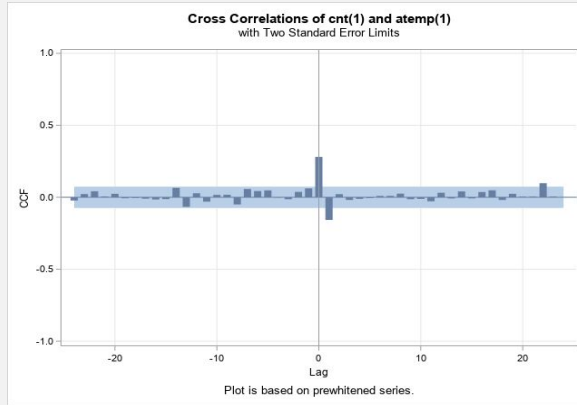


## Multivariate Models - Prewhitening Inputs

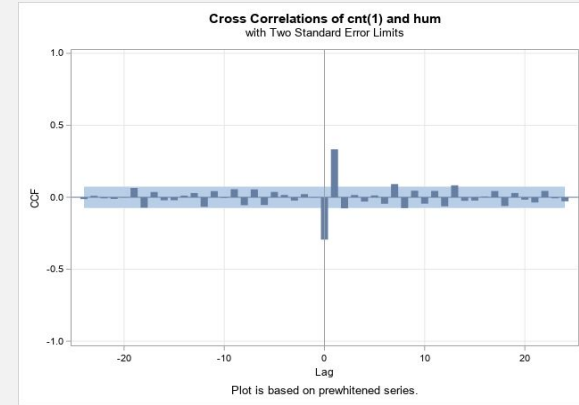
Input Variables	Prewhitening Process
ATEMP	ARIMA(1,1,2)
HUM	ARIMA(1,0,1)(0,0,1) <sub>s</sub>
WINDSPEED	Factor Model, $Q = (1,16)$
PRCP	Factor Model, $Q = (1,11)$

# Multivariate Models - Cross-correlation analysis

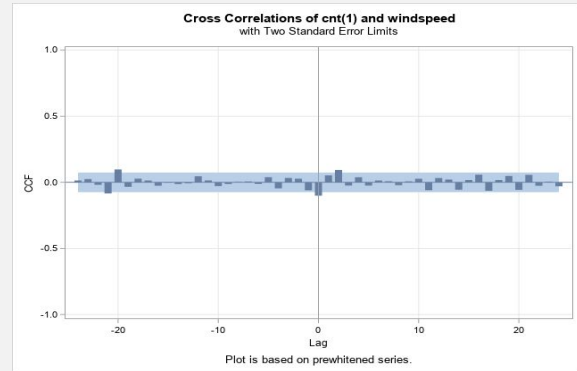
Average  
Temperature



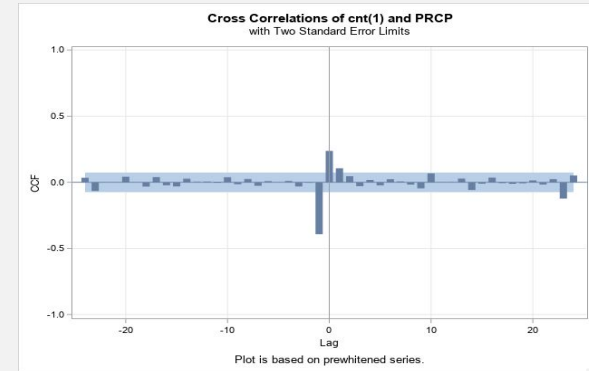
Humidity



Windspeed



Precipitation





## Multivariate Models - Estimate TF model

Prewhitened Variables	(b, r, s)
ATEMP	$b = 0, r = 0, s = 1$
HUM	$b = 0, r = 0, s = 2$
WINDSPEED	$b = 0, r = 0, s = 1$
PRCP	A significant negative lag

Exclude from TF model

# Multivariate Models - Check for adequacy(univariate)

Crosscorrelation Check of Residuals with Input atemp									
To Lag	Chi-Square	DF	Pr > ChiSq	Crosscorrelations					
5	1.27	5	0.9377	-0.005	-0.011	0.000	-0.008	-0.029	-0.026
11	3.08	11	0.9896	0.019	-0.002	0.035	-0.024	-0.012	-0.012
17	4.38	17	0.9991	0.020	0.000	0.020	0.013	0.002	0.028
23	10.45	23	0.9881	-0.008	0.003	-0.014	-0.011	0.089	-0.008
29	20.26	29	0.8846	-0.024	-0.020	0.047	-0.091	0.044	-0.006
35	33.03	35	0.5634	0.001	-0.066	0.100	-0.019	-0.053	0.012
41	36.76	41	0.6595	0.011	-0.038	0.023	0.042	-0.033	-0.016
47	40.29	47	0.7448	-0.001	0.038	0.029	-0.032	0.031	-0.022

Crosscorrelation Check of Residuals with Input hum									
To Lag	Chi-Square	DF	Pr > ChiSq	Crosscorrelations					
5	2.28	4	0.6852	-0.009	0.007	-0.001	0.018	-0.050	0.013
11	9.82	10	0.4562	-0.026	0.072	-0.057	0.016	-0.012	0.028
17	12.76	16	0.6903	-0.035	0.046	0.003	-0.006	0.016	0.020
23	18.42	22	0.6810	-0.044	-0.003	-0.004	-0.065	0.035	0.019
29	21.70	28	0.7948	-0.056	0.013	0.002	0.001	0.021	-0.027
35	26.92	34	0.8008	0.004	-0.003	0.027	-0.062	0.047	0.019
41	37.83	40	0.5684	-0.028	-0.019	0.036	-0.049	0.082	-0.059
47	41.98	46	0.6414	0.003	0.005	0.030	0.010	0.020	-0.065

Crosscorrelation Check of Residuals with Input windspeed									
To Lag	Chi-Square	DF	Pr > ChiSq	Crosscorrelations					
5	9.34	5	0.0963	0.000	-0.026	0.101	0.010	0.043	-0.002
11	12.81	11	0.3059	0.005	0.008	-0.028	0.005	0.043	-0.045
17	22.63	17	0.1616	-0.000	0.047	-0.049	-0.004	0.073	-0.059
23	29.60	23	0.1612	0.016	0.064	-0.050	0.049	-0.009	-0.010
29	38.64	29	0.1088	-0.031	0.100	-0.027	-0.008	0.019	0.018
35	41.37	35	0.2123	0.007	0.050	-0.018	-0.004	-0.025	-0.014
41	44.09	41	0.3424	0.037	0.013	-0.022	0.009	-0.000	-0.041
47	47.40	47	0.4563	0.015	0.045	-0.038	0.025	0.006	0.014

P-value are larger than 0.05



Our target and input series are not cross-correlated



TF model is appropriate

# Multivariate Models - Check for adequacy(multivariate)

Crosscorrelation Check of Residuals with Input atemp									
To Lag	Chi-Square	DF	Pr > ChiSq	Crosscorrelations					
5	1.35	5	0.9301	-0.011	-0.008	-0.025	0.000	-0.026	-0.019
11	4.89	11	0.9363	0.038	0.040	-0.001	-0.010	-0.041	-0.002
17	7.82	17	0.9703	0.027	-0.015	0.039	0.011	0.034	0.016
23	15.10	23	0.8910	-0.026	0.011	-0.032	0.025	0.084	0.024
29	30.17	29	0.4054	-0.070	-0.011	0.053	-0.093	0.059	-0.028
35	37.96	35	0.3358	0.015	-0.054	0.076	-0.011	-0.037	0.018
41	39.55	41	0.5352	-0.011	-0.018	0.029	0.018	-0.023	0.000
47	42.54	47	0.6575	-0.007	0.045	0.022	-0.012	0.032	-0.021

Crosscorrelation Check of Residuals with Input windspeed									
To Lag	Chi-Square	DF	Pr > ChiSq	Crosscorrelations					
5	20.76	5	0.0009	0.010	-0.047	0.146	-0.003	0.069	0.014
11	24.26	11	0.0117	0.000	0.004	-0.027	0.031	0.029	-0.048
17	28.89	17	0.0355	0.045	0.021	-0.031	0.004	0.047	-0.028
23	33.41	23	0.0743	0.028	0.054	-0.032	0.038	-0.004	-0.005
29	45.45	29	0.0266	-0.039	0.110	-0.031	-0.007	-0.020	0.040
35	50.63	35	0.0424	-0.012	0.070	-0.006	-0.011	0.005	-0.043
41	55.11	41	0.0694	0.023	0.041	-0.032	0.010	0.010	-0.052
47	59.33	47	0.1070	0.040	0.017	-0.036	0.034	-0.000	0.038

Crosscorrelation Check of Residuals with Input hum									
To Lag	Chi-Square	DF	Pr > ChiSq	Crosscorrelations					
5	3.13	4	0.5360	-0.011	0.007	0.003	0.019	-0.060	-0.013
11	13.68	10	0.1880	-0.036	0.092	-0.057	0.029	-0.020	0.019
17	16.55	16	0.4150	-0.051	0.021	-0.001	0.004	-0.011	0.027
23	22.04	22	0.4577	-0.052	-0.010	0.003	-0.053	0.044	-0.001
29	25.09	28	0.6228	-0.047	0.017	-0.022	0.016	0.010	-0.029
35	32.75	34	0.5286	0.018	-0.021	0.034	-0.080	0.045	0.012
41	48.83	40	0.1596	-0.057	-0.010	0.047	-0.066	0.078	-0.078
47	51.79	46	0.2582	0.032	0.020	0.008	-0.031	0.013	-0.037

P-value are larger than 0.05

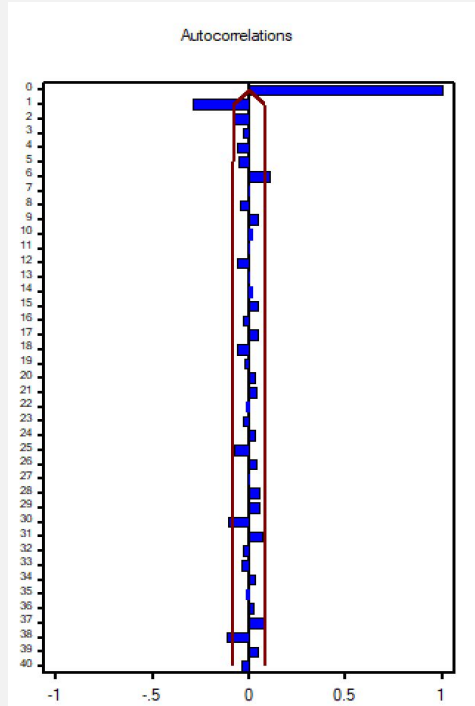


Our target and input series are not cross-correlated



TF model is appropriate

# Multivariate Models - TF model



TF model

Not WN,  
Try different  
error models

TF + ARIMA(0,1,1)

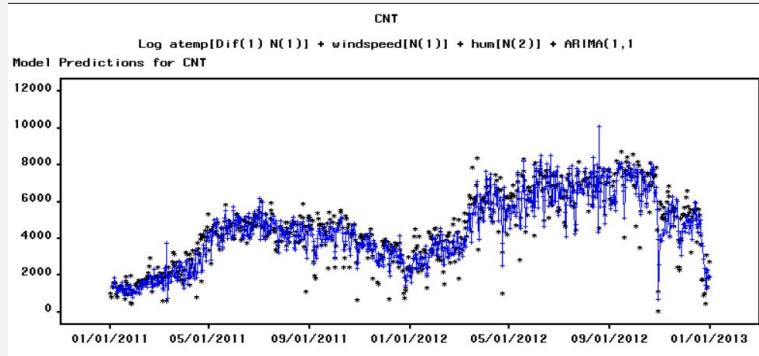
Statistics of Fit	
CNT	
Log atemp[Dif(1) N(1)] + windspeed[N(1)] + hum[N(2)] + IMA(1,1) NOINT	
Statistic of Fit	Value
Mean Square Error	1511930.6
Root Mean Square Error	1229.6
Mean Absolute Percent Error	210.73422
Mean Absolute Error	921.67957
R-Square	0.607

TF + ARIMA(1,1,1)

Statistics of Fit	
CNT	
Log atemp[Dif(1) N(1)] + windspeed[N(1)] + hum[N(2)] + ARIMA(1,1,1) NOINT	
Statistic of Fit	Value
Mean Square Error	1242800.8
Root Mean Square Error	1114.8
Mean Absolute Percent Error	202.25135
Mean Absolute Error	857.75356
R-Square	0.677

Comparison of TF-noise model

# Multivariate Models - TF-noise model



Parameter Estimates

CNT

Log atemp[Dif(1) N(1)] + windspeed[N(1)] + hum[N(2)] + ARIMA(1,1,1) NOINT

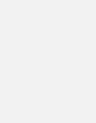
Model Parameter	Estimate	Std. Error	T	Prob> T
Moving Average, Lag 1	0.84890	0.0275	30.8383	<.0001
Autoregressive, Lag 1	0.20554	0.0505	4.0682	0.0001
ATEMP[Dif(1) N(1)]	1.08876	0.1513	7.1942	<.0001
ATEMP[Dif(1) N(1)] Num1	-0.39755	0.1473	-2.6984	0.0083
WINDSPEED[N(1)]	-0.86002	0.1338	-6.4267	<.0001
WINDSPEED[N(1)] Num1	0.49965	0.1425	3.5053	0.0007
HUM[N(2)]	-1.00334	0.0794	-12.6372	<.0001
HUM[N(2)] Num1	-0.18655	0.0783	-2.3819	0.0193
HUM[N(2)] Num2	-0.11360	0.0765	-1.4854	0.1409
Model Variance (sigma squared)	0.05477	.	.	.

- Stationary
- Not WN
- **RMSE = 1114.8**
- **MAPE = 202.25135**



Model has been improved





## Part 4

# TF Model with Regressors



# TF Model with regressors - Regressors Detection

## Feedback Relationships with PRCP

- **Significant Regressors:**  
PRCP, PRCP of Tomorrow( $k = -1$ )
- **Interpretation:**  
People can check tomorrow's weather. The decision today will depend on precipitation volumes of yesterday, today and tomorrow.  
  
Before the inclement weather, the level of bike trips count is high.

## Holiday Effect

Holiday	2011		2012	
Martin Luther King Jr. Day	Jan. 15th	Jan. 18th	Jan. 14th	Jan. 17th
Valentine's Day		Feb. 14th		Feb. 14th
Presidents' Day	Feb. 19th	Feb. 22th	Feb. 18th	Feb. 21th
Labor Day	Sep. 3rd	Sep. 6th	Sep. 1st	Sep. 4th
Thanksgiving Day	Nov. 24th	Nov. 25th	Nov. 22th	Nov. 23th
Black Friday		Nov. 26th		Nov. 23th
Christmas Eve	Dec. 24th		Dec. 24th	

### Outliers:

Notable deviation substantially higher or lower than predicted values in former ARIMA or TF models

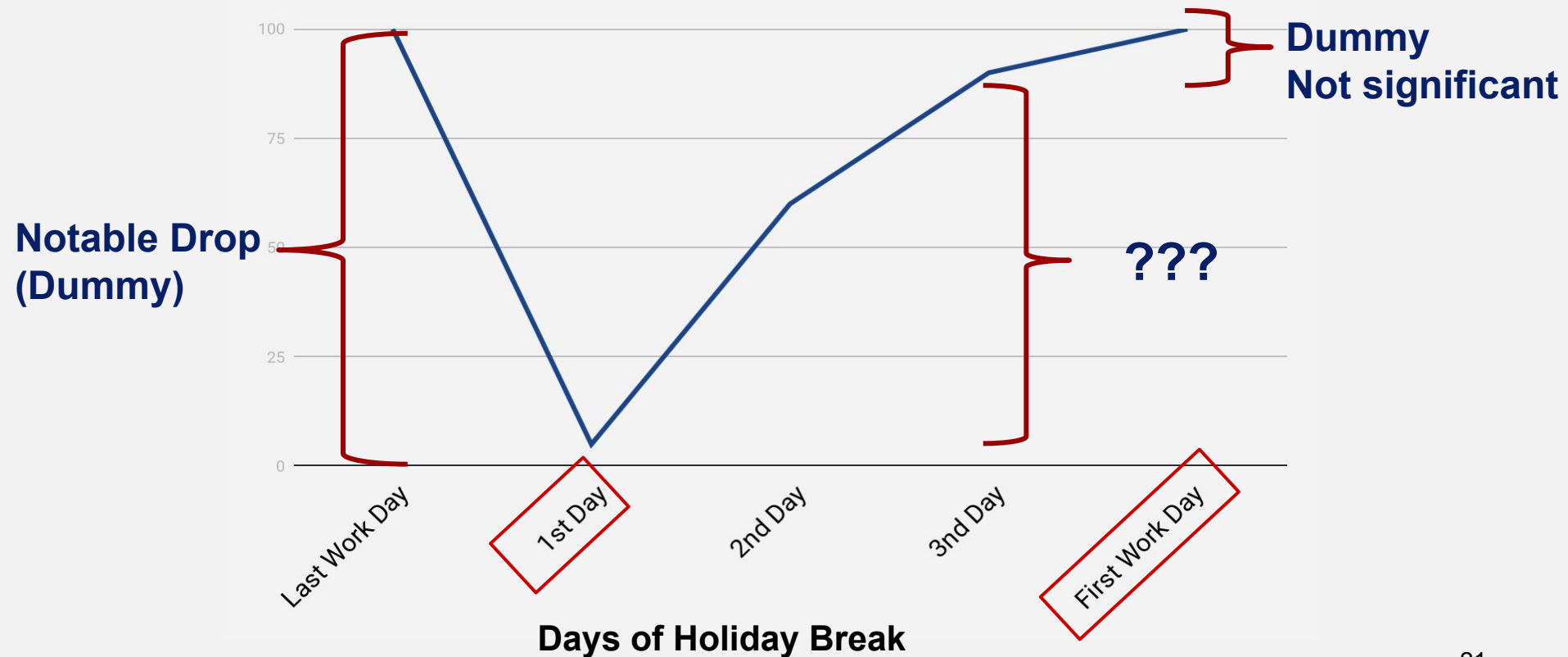
### Dummy Variables:

Break start days (Significantly Negative)

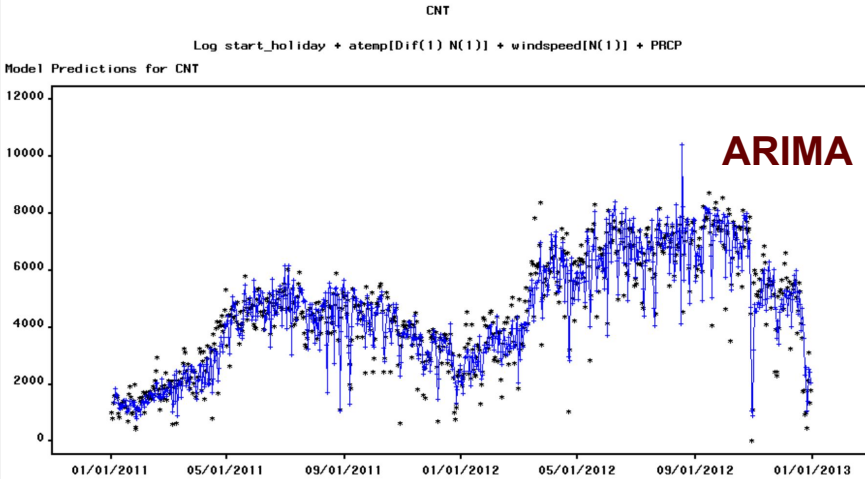
The first working day (Not Significant)



## TF Model with regressors - Count of Bike Trips during Holiday



# TF Model with regressors - Model Identification and Comparison

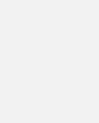


Parameter Estimates				
CNT				
Log start_holiday + atemp[Dif(1) N(1)] + windspeed[N(1)] + PRCP + PRCP_LAG + hum[N(2)] +				
Model Parameter	Estimate	Std. Error	T	Prob> T
Intercept	0.00283	0.0016	1.7485	0.0839
Moving Average, Lag 1	0.85473	0.0274	31.2373	<.0001
Autoregressive, Lag 1	0.24369	0.0508	4.7986	<.0001
start_holiday	-0.21953	0.0688	-3.1915	0.0020
ATEMP[Dif(1) N(1)]	1.01249	0.1352	7.4876	<.0001
ATEMP[Dif(1) N(1)] Num1	-0.40382	0.1318	-3.0632	0.0029
WINDSPEED[N(1)]	-0.61818	0.1222	-5.0581	<.0001
WINDSPEED[N(1)] Num1	0.33574	0.1299	2.5855	0.0114
PRCP	-0.05024	0.0244	-2.0563	0.0428
PRCP_LAG	-0.29191	0.0243	-12.0110	<.0001
HUM[N(2)]	-0.69635	0.0767	-9.0834	<.0001
HUM[N(2)] Num1	-0.18074	0.0730	-2.4753	0.0153
HUM[N(2)] Num2	-0.13229	0.0686	-1.9274	0.0572
Model Variance (sigma squared)	0.04379	.	.	.

TF

Regressors

	TF Model	TF Model with regressor
Root Mean Square Error	1114.8	891.87
Mean Absolute Percent Error	202.25	68.99



# Part 5

## Vector Model

# Vector Model

## Cross Correlations

Schematic Representation of Cross Correlations													
Variable/Lag	0	1	2	3	4	5	6	7	8	9	10	11	12
cnt_delta_std	++	--	-.	-.	..	..	+.	..	..	..	..	..	..
PRCP1	++	++	..	..	..	..	..	..	..	..	..	+.	..
+ is > 2*std error, - is < -2*std error, . is between													

## Partial Autoregression

Schematic Representation of Partial Autoregression												
Variable/Lag	1	2	3	4	5	6	7	8	9	10	11	12
cnt_delta_std	-.	-.	-.	-.	-.	+.	..	..	..	..	..	..
PRCP1	-.	-.	-.	-.	-.	-.	-.	..	..	..	+.	..
+ is $> 2 \times \text{std error}$ , - is $< -2 \times \text{std error}$ , . is between												

## VMA(2) Parameter Estimates

Model Parameter Estimates						
Equation	Parameter	Estimate	Standard Error	t Value	Pr >  t	Variable
cnt_delta_std	MA1_1_1	0.55096	0.03353	16.43	0.0001	e1(t-1)
	MA1_1_2	-0.05861	0.03596	-1.63	0.1036	e2(t-1)
	MA2_1_1	0.25936	0.03202	8.10	0.0001	e1(t-2)
	MA2_1_2	-0.00144	0.03470	-0.04	0.9669	e2(t-2)
PRCP1	MA1_2_1	0.55967	0.03711	15.08	0.0001	e1(t-1)
	MA1_2_2	-0.09792	0.03719	-2.63	0.0086	e2(t-1)
	MA2_2_1	0.06901	0.04136	1.67	0.0957	e1(t-2)
	MA2_2_2	-0.03966	0.04289	-0.92	0.3555	e2(t-2)

# Vector Model

## Matrix Function

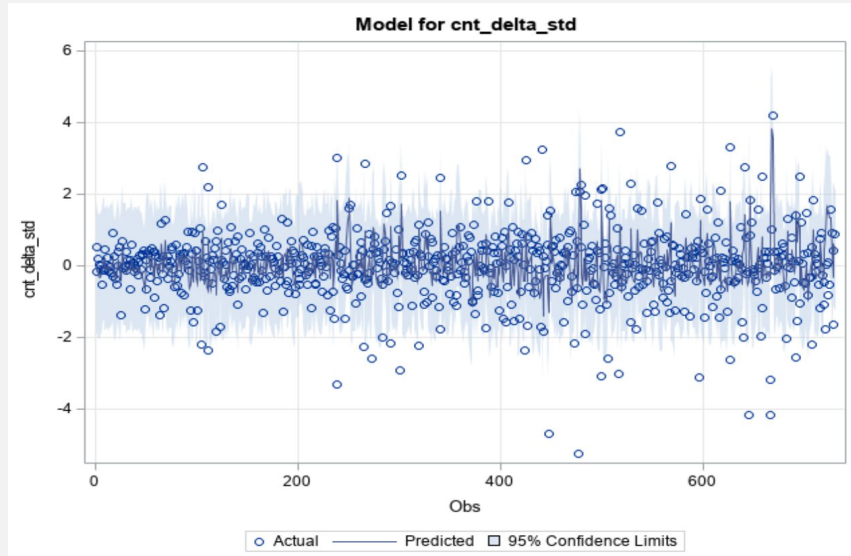
$$\begin{bmatrix} \Delta CNT\_STD \\ PRCP\_STD \end{bmatrix} = \begin{bmatrix} e_{1t} \\ e_{2t} \end{bmatrix} - \begin{bmatrix} 0.55096 & -0.05861 \\ 0.55967 & -0.09792 \end{bmatrix} \times \begin{bmatrix} e_{1t-1} \\ e_{2t-1} \end{bmatrix} - \begin{bmatrix} 0.25936 & -0.00144 \\ 0.06901 & -0.03966 \end{bmatrix} \times \begin{bmatrix} e_{1t-2} \\ e_{2t-2} \end{bmatrix}$$

## Correlations of Residuals

Schematic Representation of Cross Correlations

Variable/Lag	0	1	2	3	4	5	6	7	8	9	10	11	12
cnt_delta_std	+. .	..	..	-.	..	..	+. .	..	..	..	..	..	..
PRCP1	+. .	..	..	..	..	..	+. .	..	..	..	..	+. .	..

+ is > 2\*std error, - is < -2\*std error, . is be





**Thank you for listening !**