

# Demand Forecasting to Increase Profits on Perishable Items



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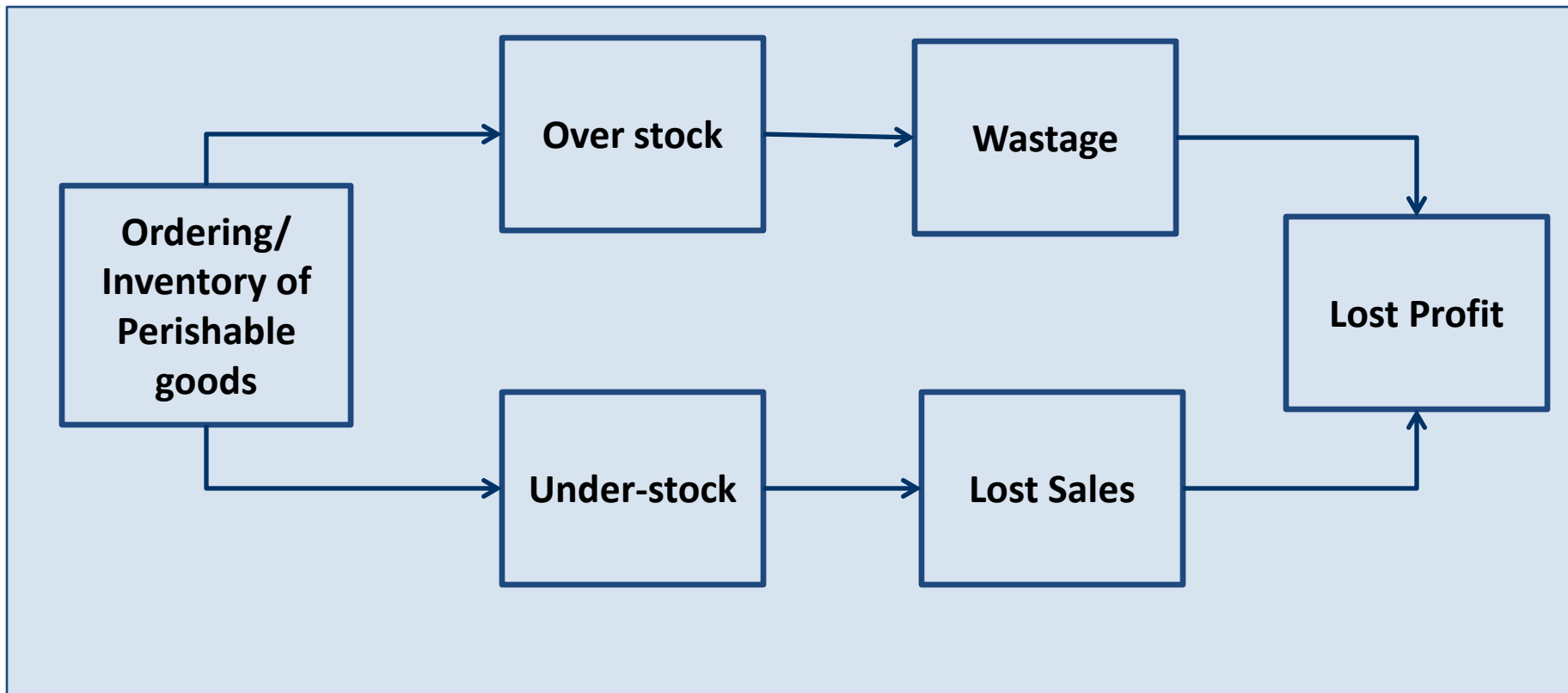
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**Sharath Srinivas**

**Forecasting Analytics, ISB**



**GOAL: Maximize profits when selling perishable items such as fruits and vegetables**

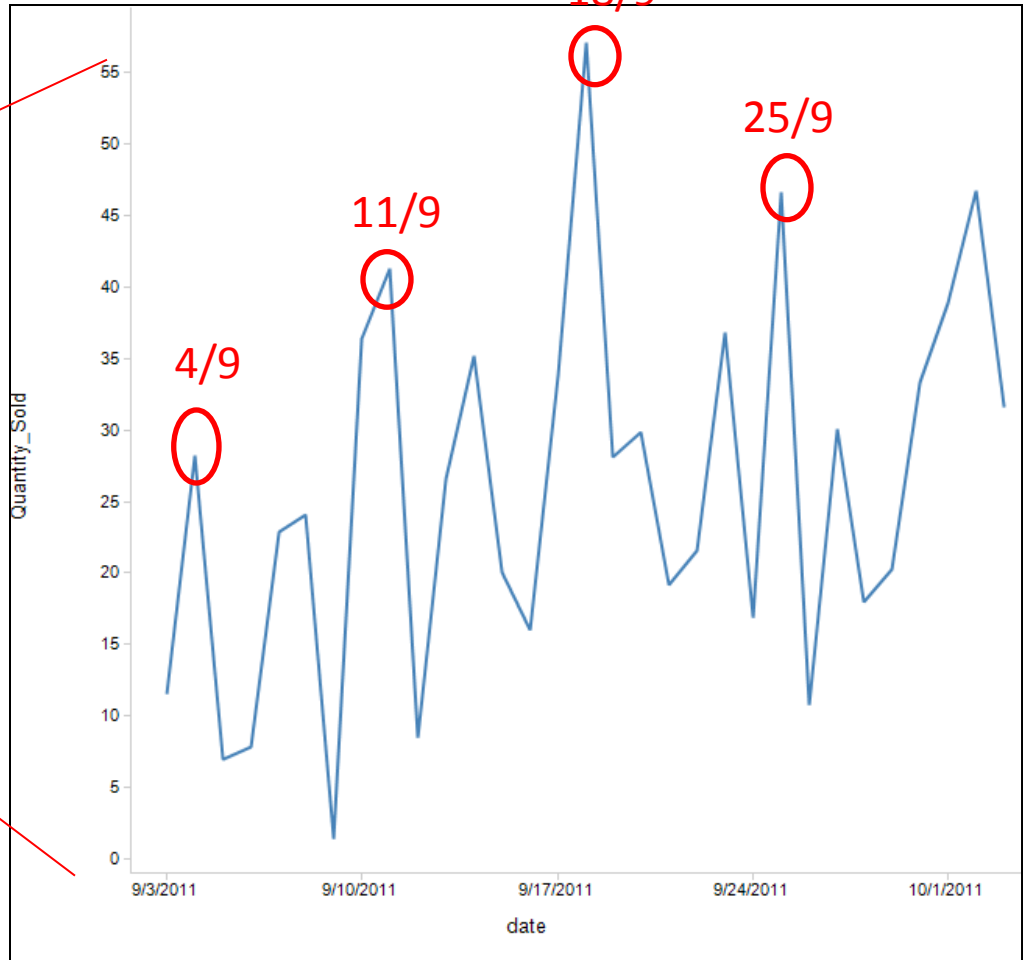
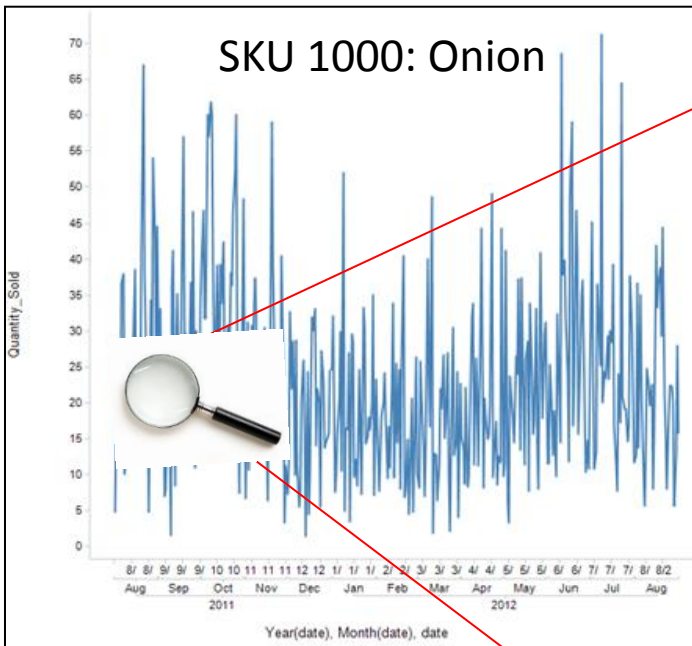


## Data:

- Hypermart customer Transaction data from 8/1/2011 to 8/31/2012
- Each transaction includes the customer ID, SKU and purchase quantity
- 5 SKUs were explored: Banana, Apple, Onion, Tomato, Papaya

## Caveats:

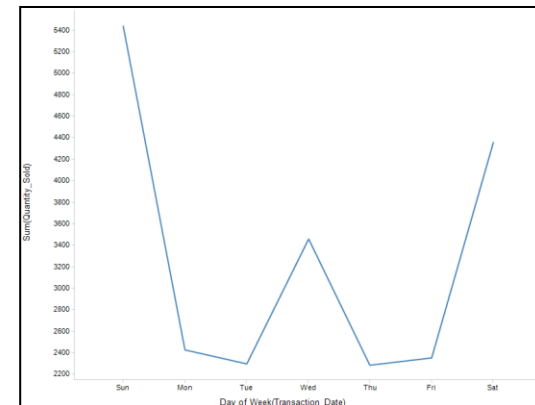
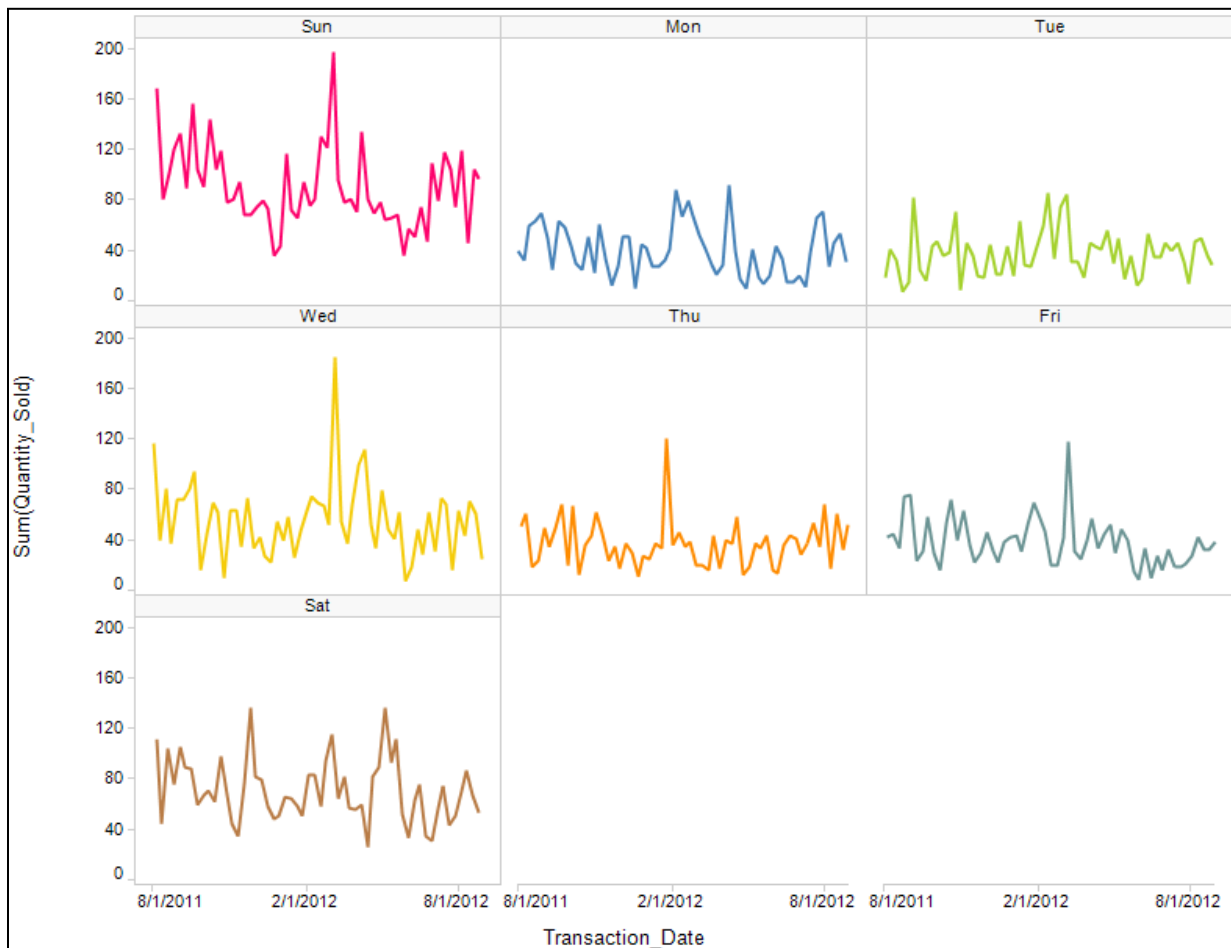
- Transaction data includes only loyalty card purchases
- Data does not include promotions
- Data includes only customer demand and not indicate inventory levels, procurement etc.
- Infrequent visits by customers to the store



### September

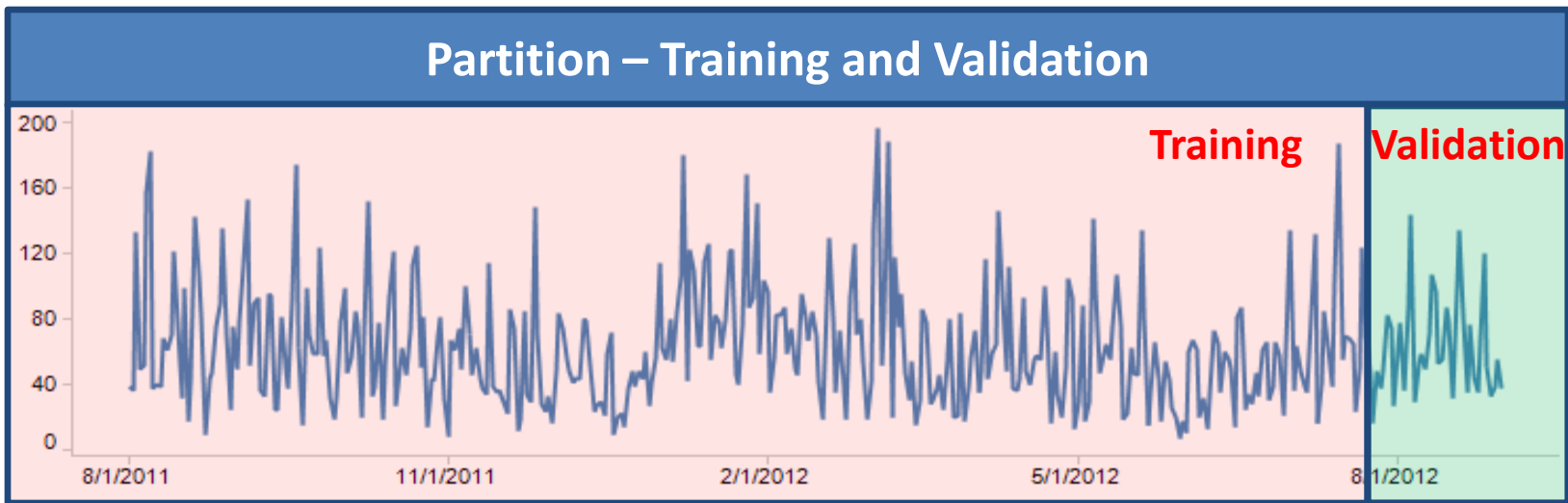
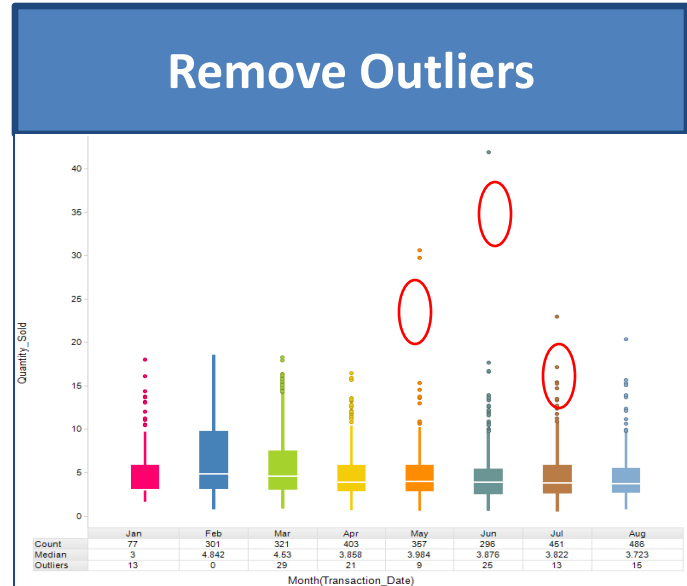
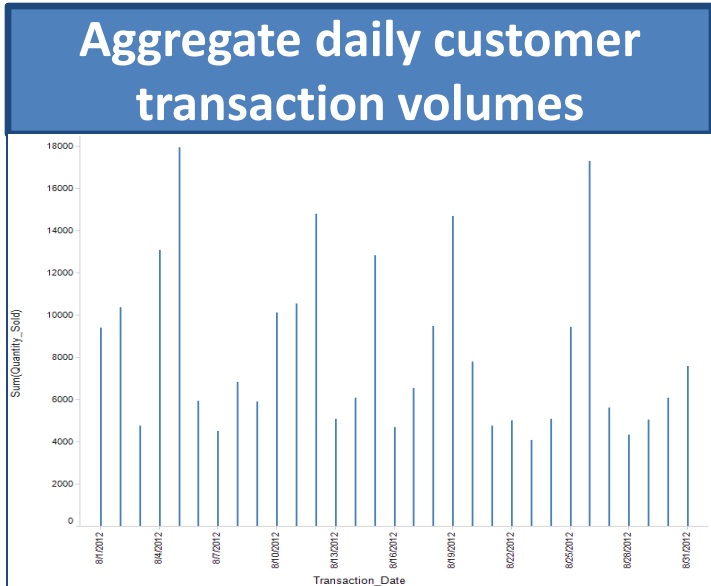
S	M	T	W	T	F	S
				1	2	3
4	5	6	7	8	9	10
11	12	13	14	15	16	17
18	19	20	21	22	23	24
25	26	27	28	29	30	

**Hypothesis: There is weekly seasonality**



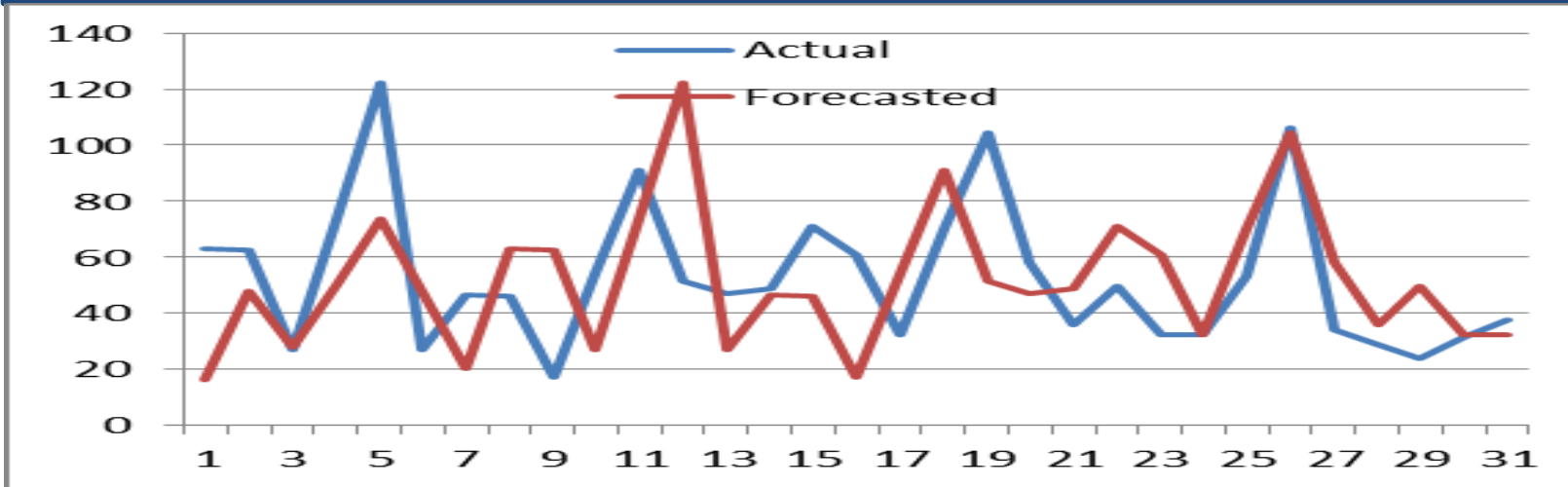
**Higher demand on:**

- Sunday
- Saturday
- Wednesday





### Naïve Model: Forecast for Aug 2012

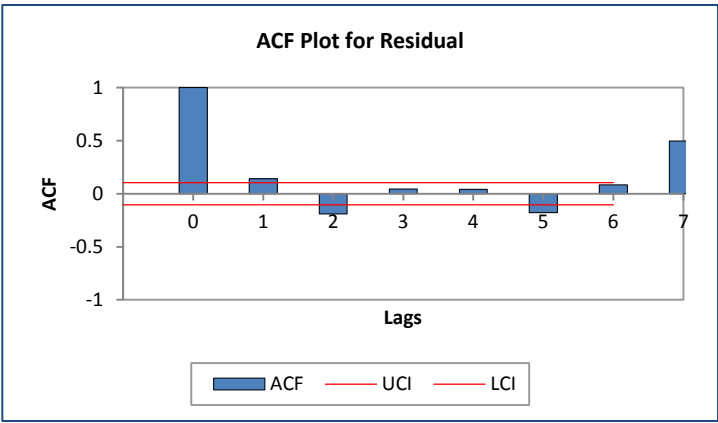
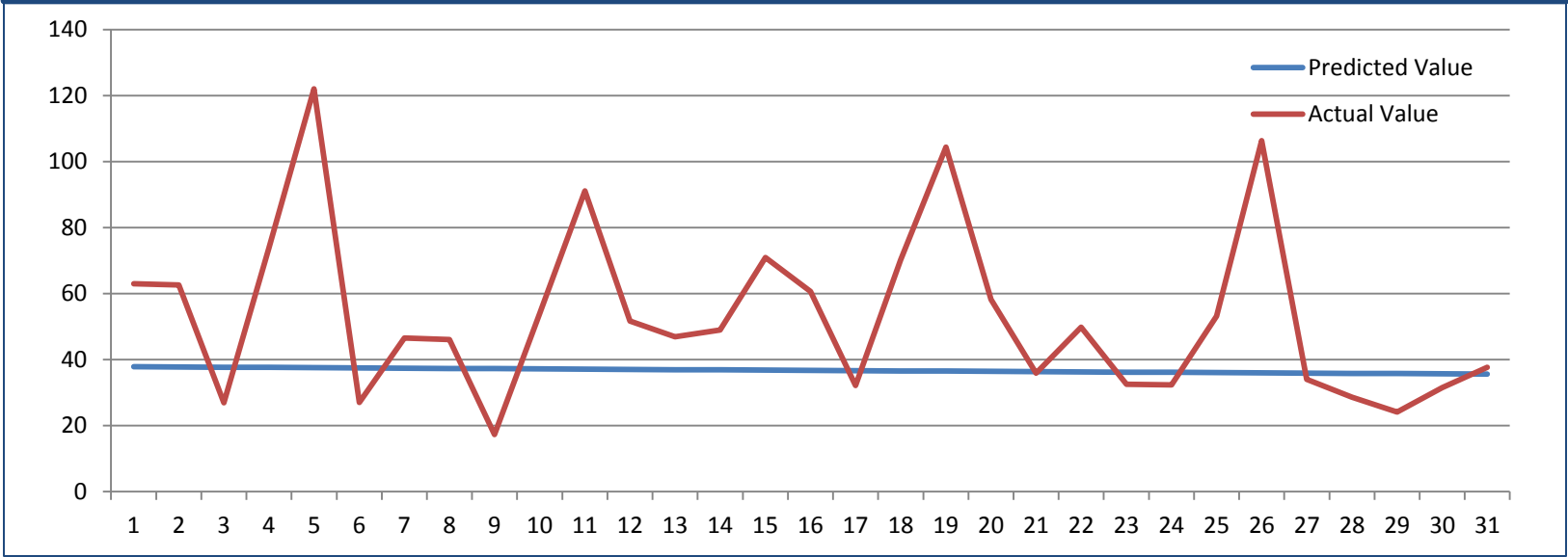


MAE	22.632
Average Error	1.004903
MAPE	50.15%
RMSE	791.4196

- Naïve model can only be used as a benchmark
- Accuracy of the model is very low



## Linear Regression Model (With no Dummy Variables): Forecast for Aug 2012



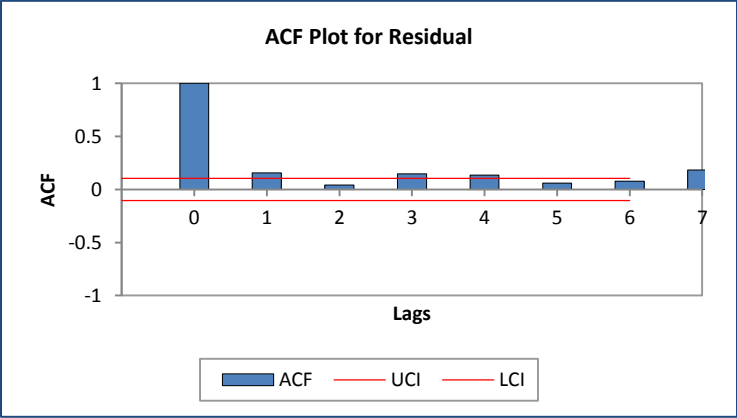
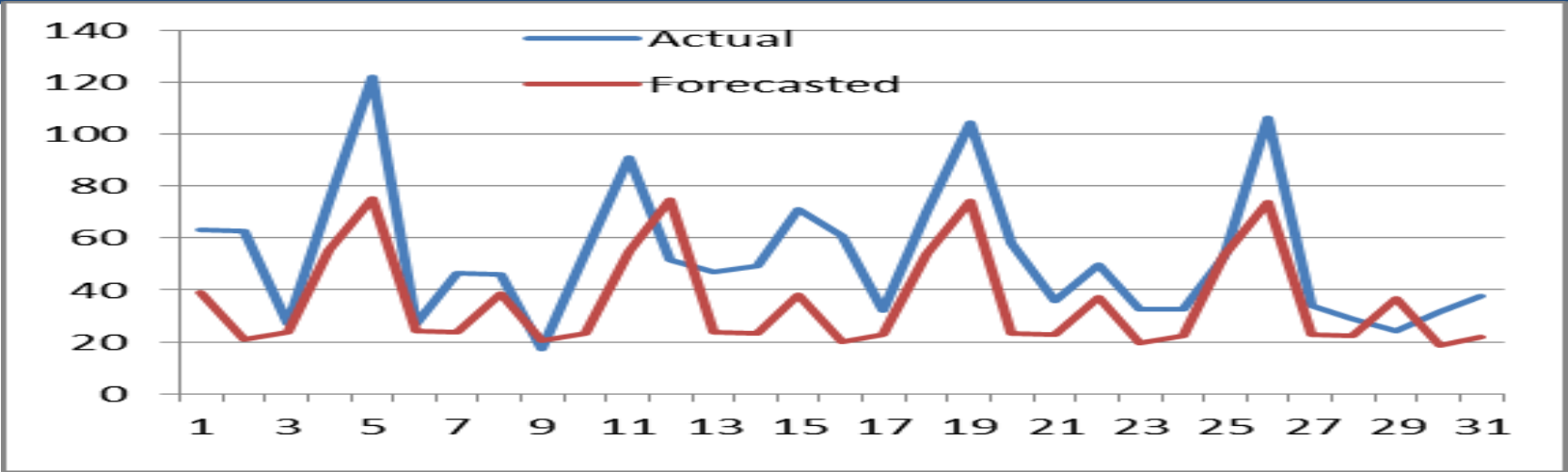
<b>MAE</b>	21.26407
<b>Average Error</b>	16.19199
<b>MAPE</b>	35.33%
<b>RMSE</b>	896.3684

- ACF Plot shows that there is seasonality left in the residuals**





## Linear Regression Model (With Dummy Variables): Forecast for Aug 2012

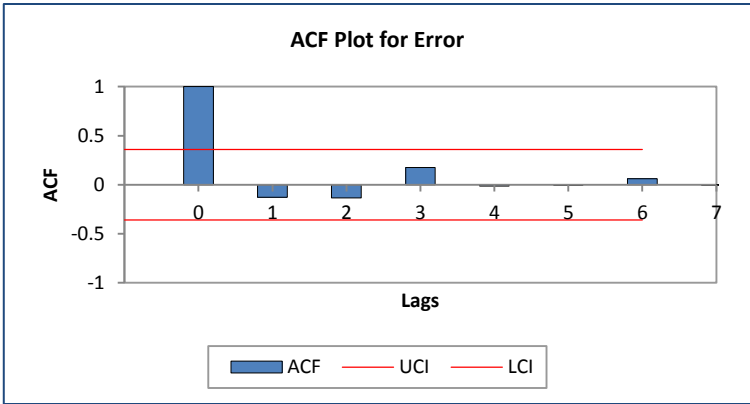
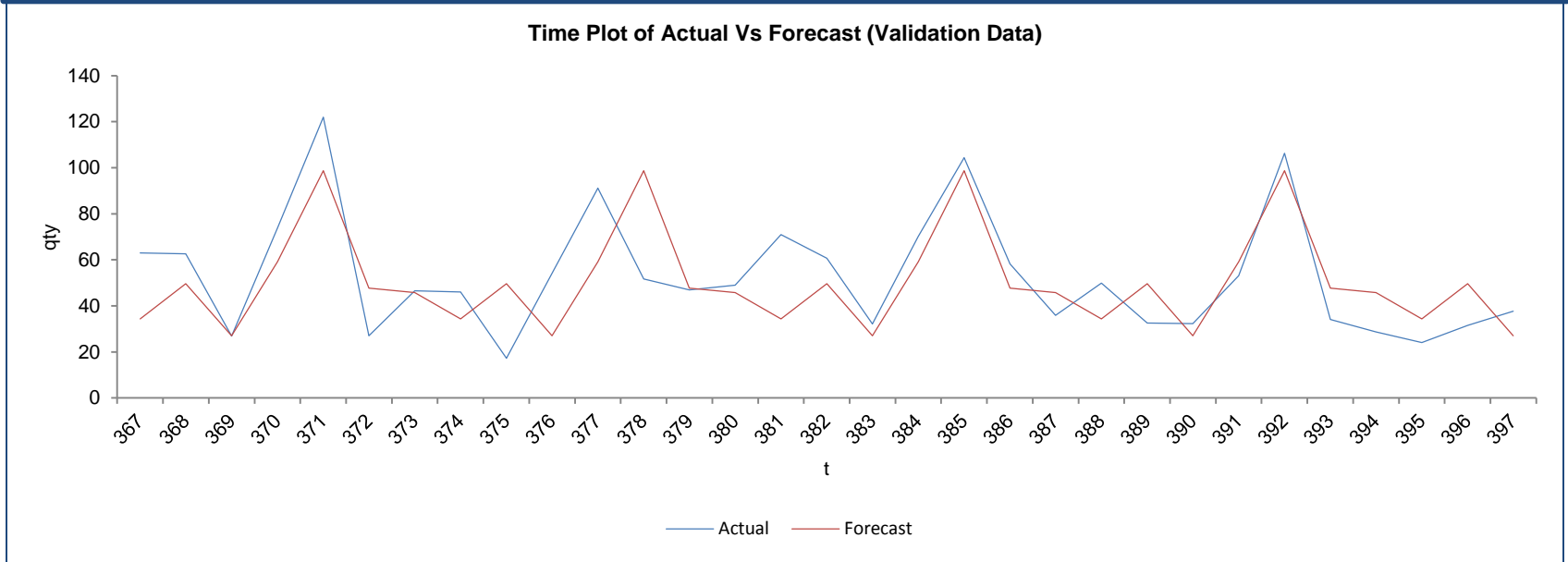


<b>MAE</b>	19.8724
<b>Average Error</b>	17.33157
<b>MAPE</b>	36.46%
<b>RMSE</b>	552.8134

- ACF Plot shows that there is seasonality left in the residuals**

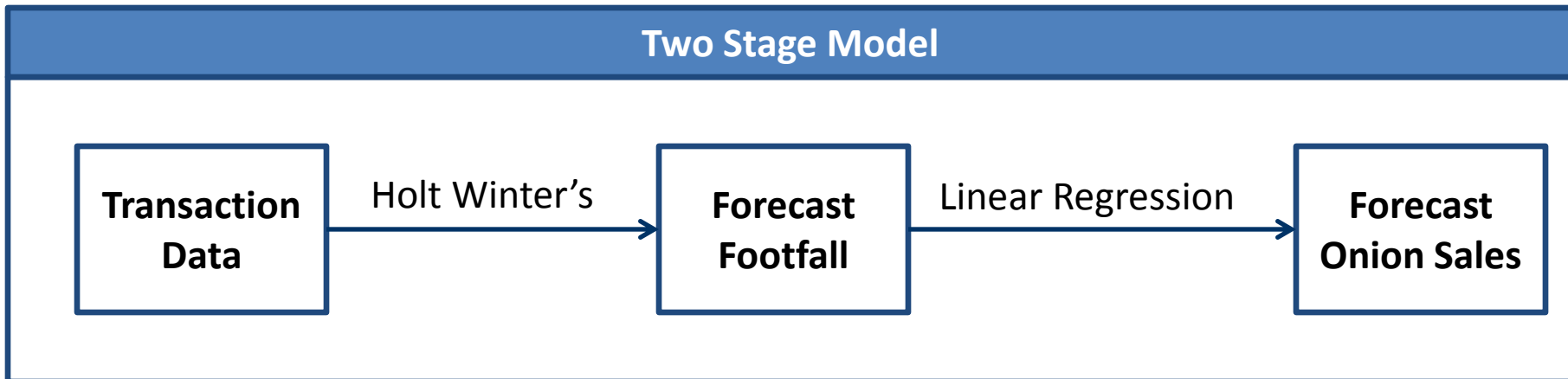


## Holt Winters: Forecast for Aug 2012



<b>MAE</b>	15.04821
<b>Average Error</b>	2.589459
<b>MAPE</b>	34.84%
<b>RMSE</b>	350.3683

Seasonality handled by Holt Winters method

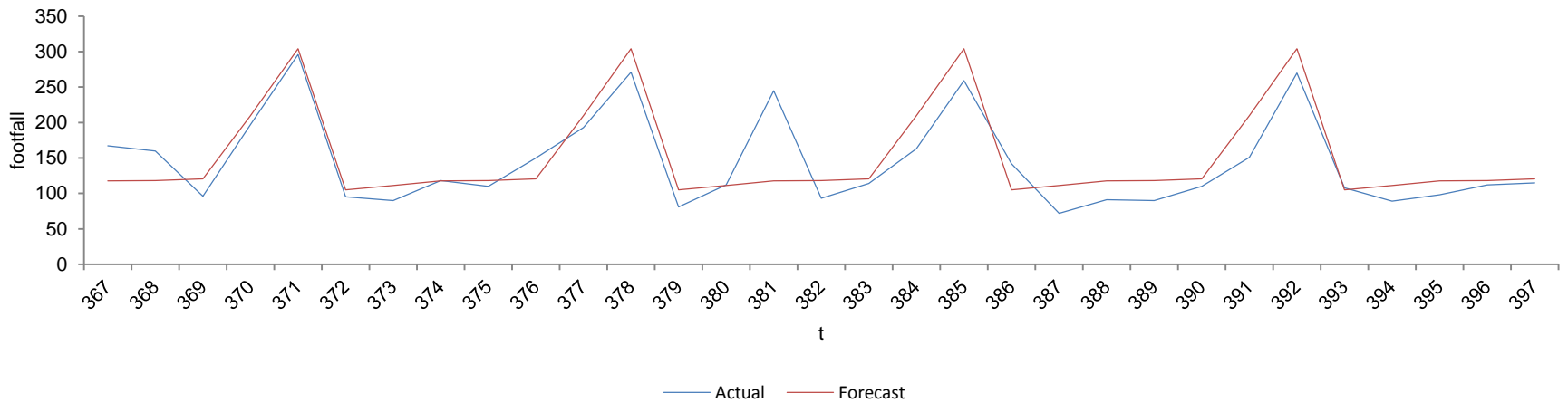


- **Sales of individual SKUs categorized on Day of the week is noisy**
- **Instead we forecast amount of Footfall in the store**
- **Use footfall as a proxy to forecast the SKU quantity demand**



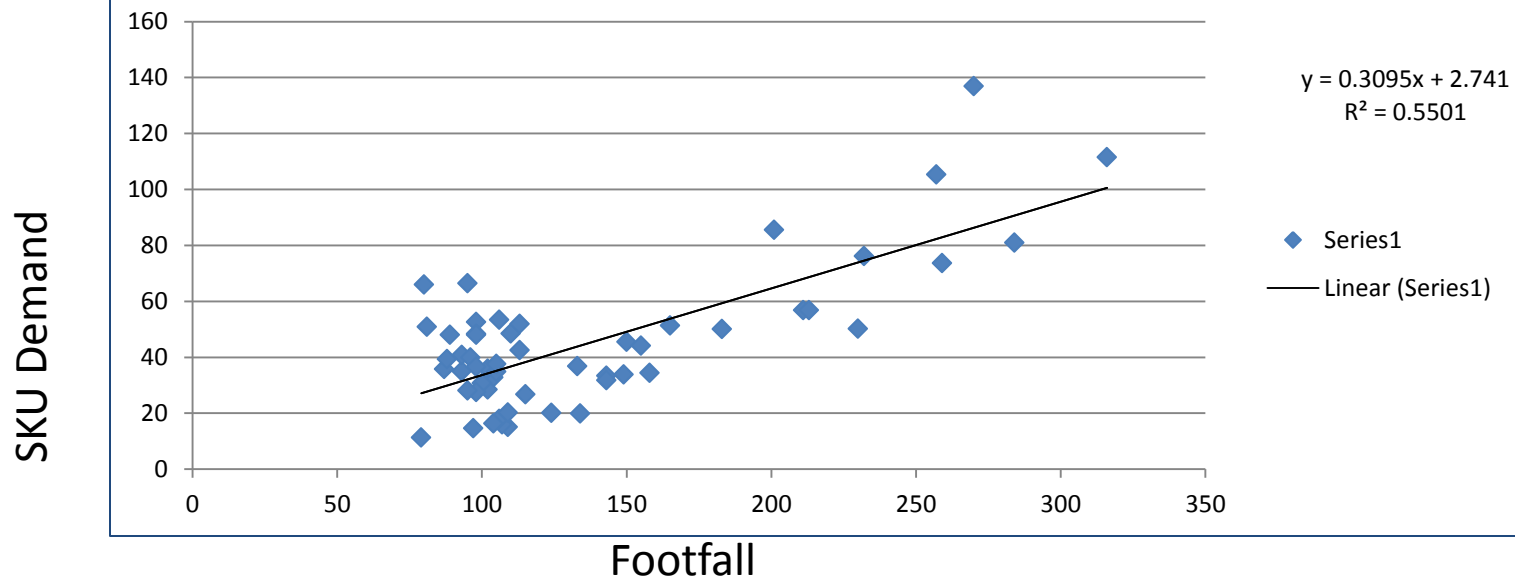
## Step 1: Forecast Footfall

Time Plot of Actual Vs Forecast (Validation Data)



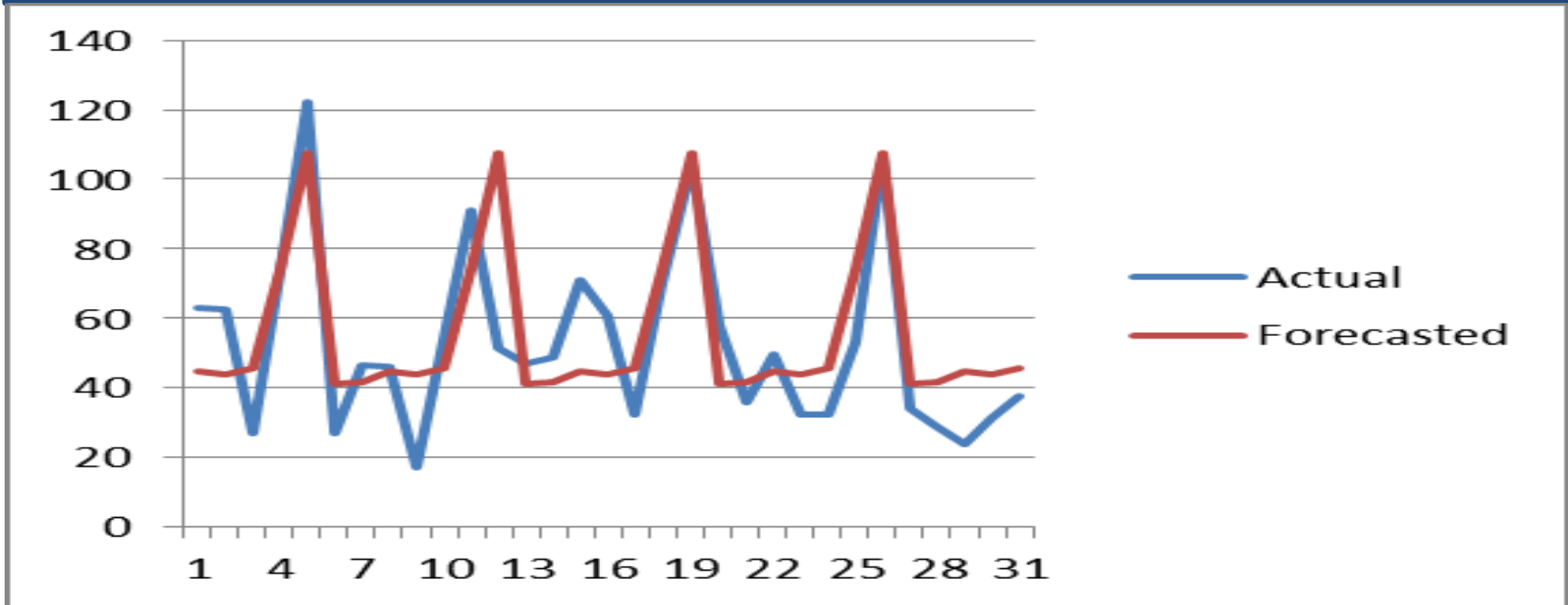


## Step 2: Forecast Sales of SKU 1000





## 2 Staged Model: Forecast for Aug 2012

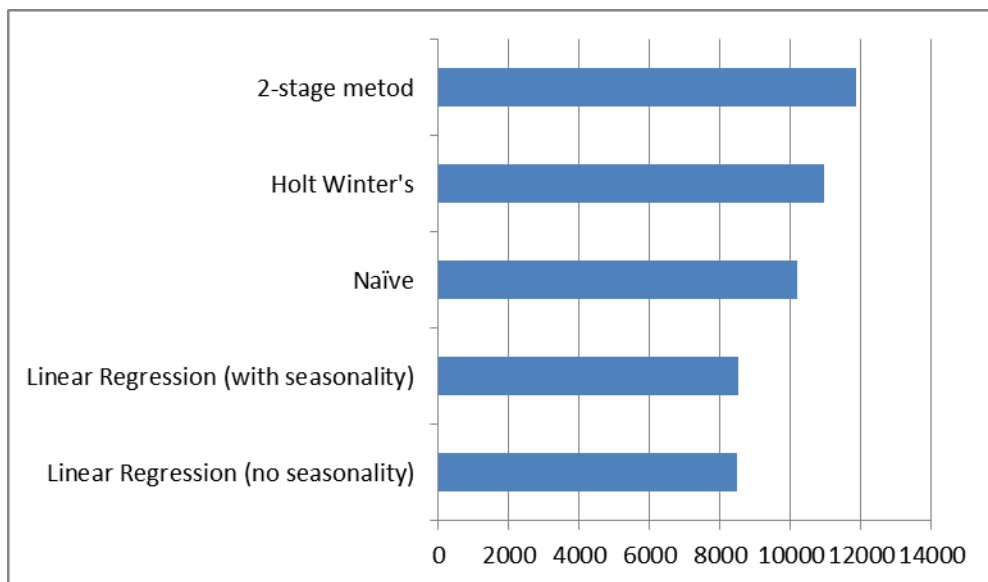


MAE	13.27165
Average Error	-2.95764
MAPE	33.33%
RMSE	286.4652



## Two step process:

1. Determine a cost-metric (e.g. profit: See Appendix for profit calculations)
2. Evaluate the effect of different forecasting methods on the metric



Achieved **17% improvement** in profitability by **leveraging** advanced forecasting techniques at SKU level (compared to baseline-naïve forecasting)

# Recommendations

- By forecasting demand at the SKU level, the store can increase profitability by:
  - Reducing wastage
  - Reducing lost sales
- Two-staged model offers best performance in terms of profitability improvement
- **Key Learning:** Handling noisy data, *“Torture the data, and it will confess to anything!”*



# Appendix

## A Model of System Mechanics

- Suppose  $Q = 100$ 
  - suppose demand turns out to be  $D = 110$
  - profit =  $\$70,000(100) - \$55,000(100) = \$1,500,000$
- Now suppose  $Q = 100$  and demand turns out to be  $D = 90$ 
  - profit =  $\$70,000(90) + \$20,000(10) - \$55,000(100) = \$1,000,000$

- The General Formula:

	Sales	Salvage	COGS
if $D \geq Q$	$P \cdot Q$	0	$C \cdot Q$
if $D < Q$	$P \cdot D$	$S \times (Q - D)$	$C \cdot Q$

- As a single formula...
  - profit( $Q, D$ ) = IF( $D < Q$ ,  $(P \times D) + S \times (Q - D) - (C \times Q)$ ,  $(P \times Q) - (C \times Q)$ )