

# Polling Patients: Elevating the Healthcare Experience

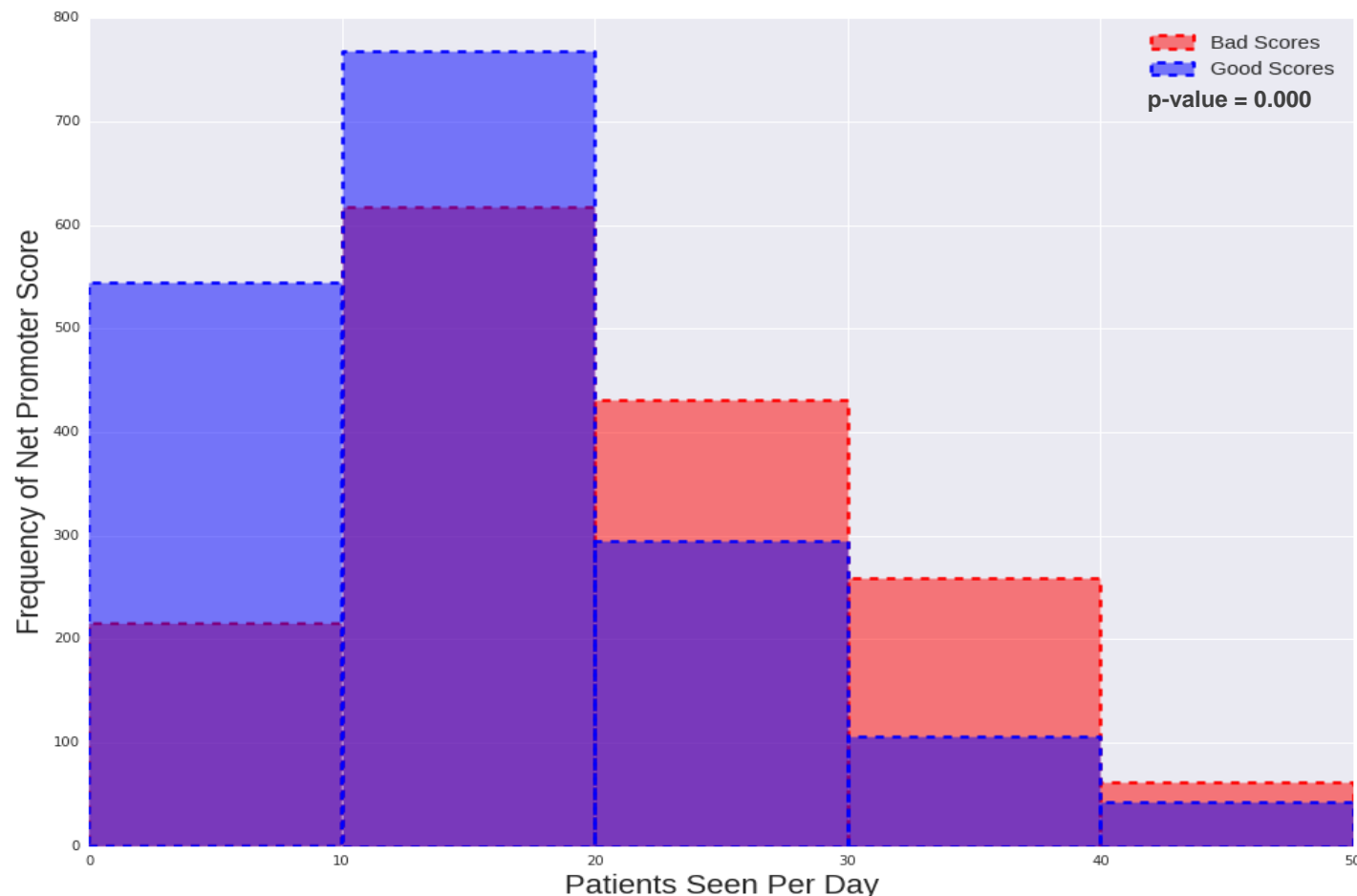
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**ANJALI SHAH**

INSIGHT PROJECT PRESENTATION

# Finding the Optimal Number of Patients Per Provider

**Objective:** Finding the optimal number of patients per provider to get good scores



## Approach

- Net Promoter Score (NPS) was binarized into good ( $\geq 90$ th percentile) and bad ( $< 90$ th percentile)
- Non-parametric Mann-Whitney U test was performed to compare if the means of good and bad score distributions were significantly different

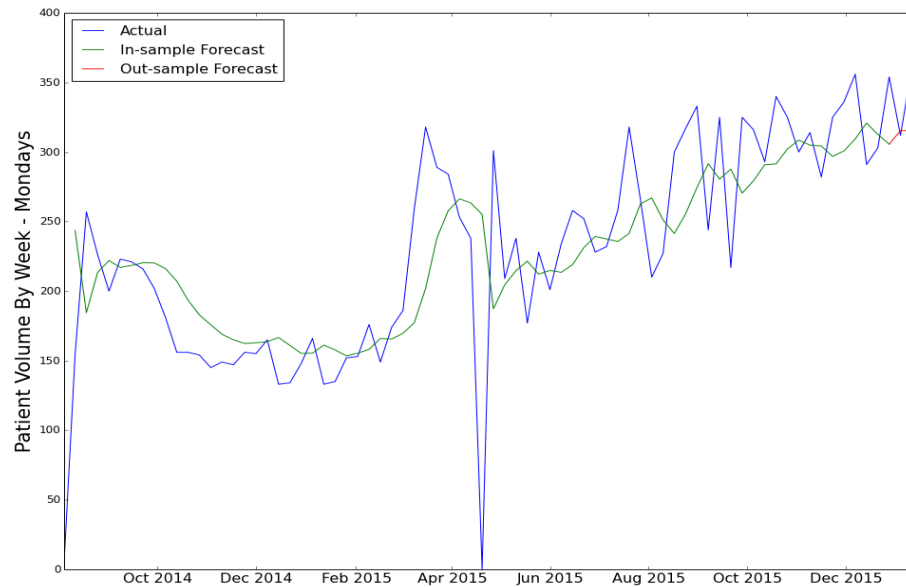
## Actionable Insights

- Statistically significant difference between good and bad score distributions (p-value = 0.000)
- Patients per provider in a day (95% CL):
  - ✓ Good Scores: 14 – 15 patients
  - ✓ Bad Scores: 19 – 20 patients

# Forecasting Patient Volume (Weekly Data)

**Objective: Predicting patient volume to determine optimal number of providers per site**

**ARIMA (p=2,d=0,q=1) Model**  
(Non-Stationary Model)



## Approach

- Used ACF and PACF plots to determine order of ARIMA model
- Applied an iterative approach to reduce prediction errors

**ARIMA (p=2,d=1,q=1) Model**  
(Stationary Model)

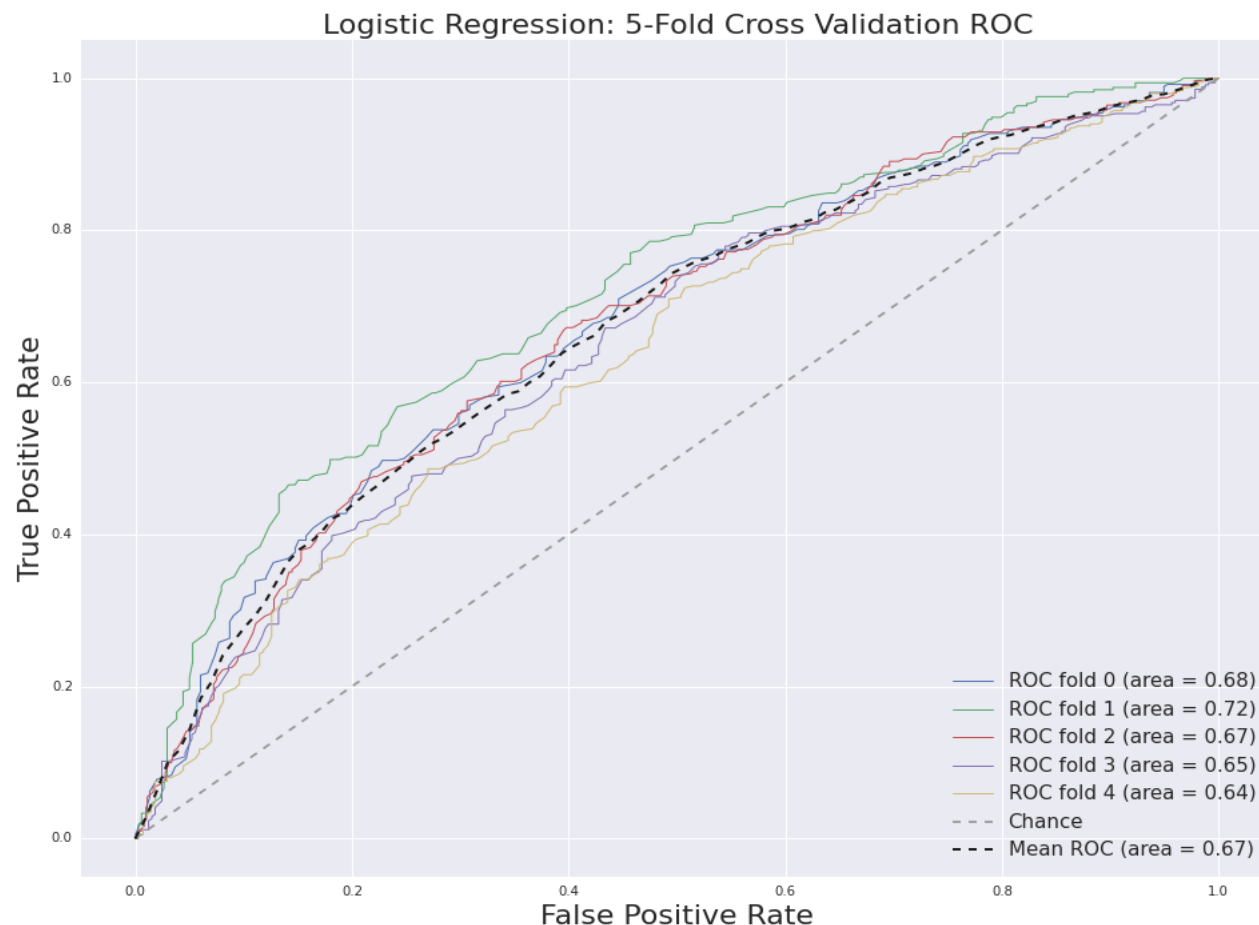


## Actionable Insights

- Mean absolute percentage error = ~17%
- Predicted volume outside sample (2016-01-18) is ~330 patients

# Classifying Scores by Weekday & Patients Per Day

**Objective:** Finding probability of good scores based on patient-per-provider and day-of-week features



## Approach

- Four different classification models were fitted to the training dataset
- Total running time and performance (ROC AUC) were measured on test dataset to determine the best model

Model	Total Running Time	ROC AUC using 10-fold CV
Logistic Regression	0.004	0.68
Naïve Bayes	0.002	0.64
Support Vector Classification (SVC)	0.16	0.67
Random Forest	0.03	0.62

**Selected model based on performance and total running time**

# Putting It All Together

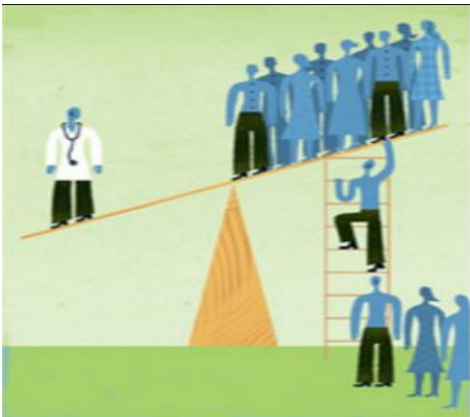
## Final Product

Statistical Mann-Whitney U Test  
Time Series Analysis  
Logistic Regression Model

<http://anjalibshah.github.io/Elevating-Healthcare-Experience/>

How will it help the startup and elevate patients' healthcare experience?

Optimum Management



Better Experience



Happy Customers



# Anjali Shah



## PhD

- Biomedical Informatics

## Masters and Bachelors

- Computer Science and Engineering

## Professional Experience

- 10+ years of professional experience across education, healthcare, financial services, and telecom sectors

### Former Employers



BNP PARIBAS



## Professional Interests / Expertise

- Worked at the intersection of data and technology to solve interesting business problems:
  - ✓ Rutgers (EHR Data)
  - ✓ BNP Paribas (Legal Data )
  - ✓ Goldman Sachs (Reference Data)
- Passionate about trends shaping healthcare industry with expertise in EHR and data analytics

## Passionate About

- Travel (widely traveled across US, Europe and Asia)
- Yoga
- Scrapbooking

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# Appendix

# Algorithm and Data Analysis Approach

## Algorithm Stages and Pipeline



## Features

- Customer ID
- Site ID
- Provider ID
- Patient ID
- Visit ID
- Date Seen
- Score

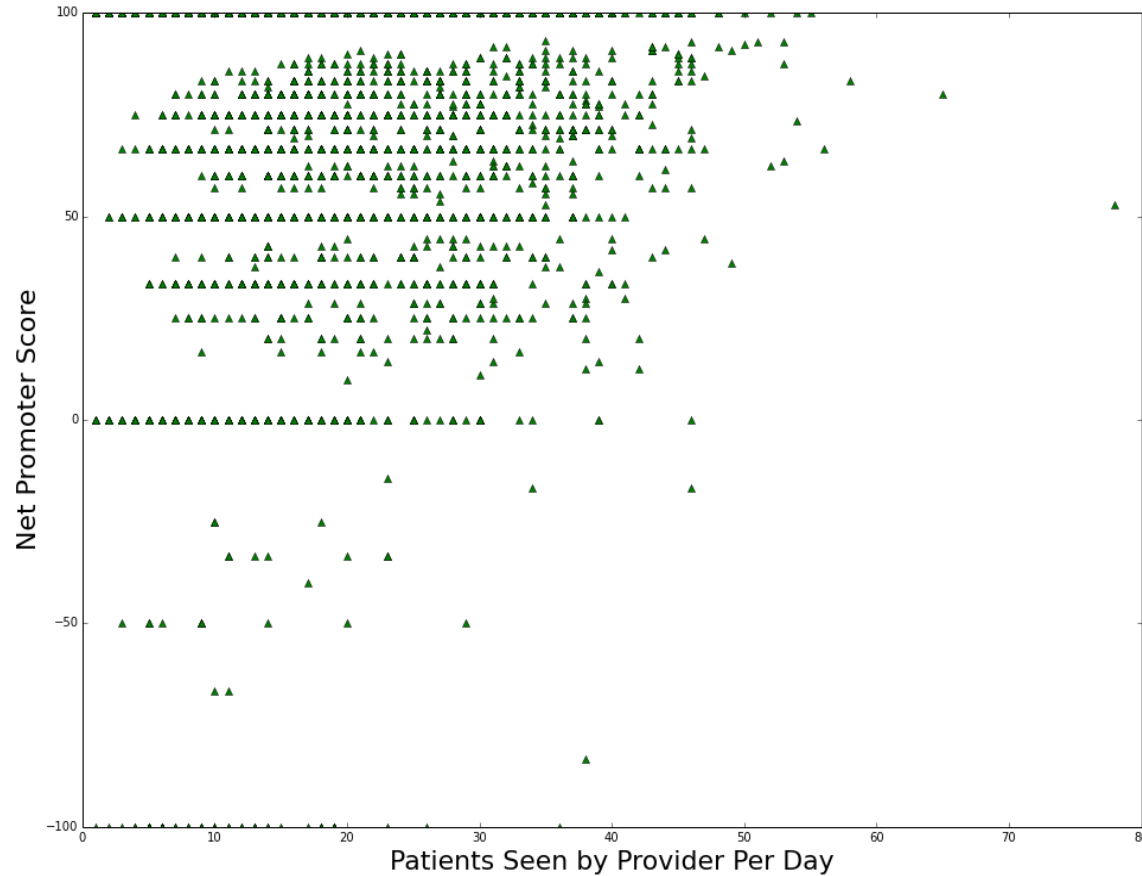


1. Patients seen by provider per day
2. Day of the week
3. Net Promoter Score (NPS)

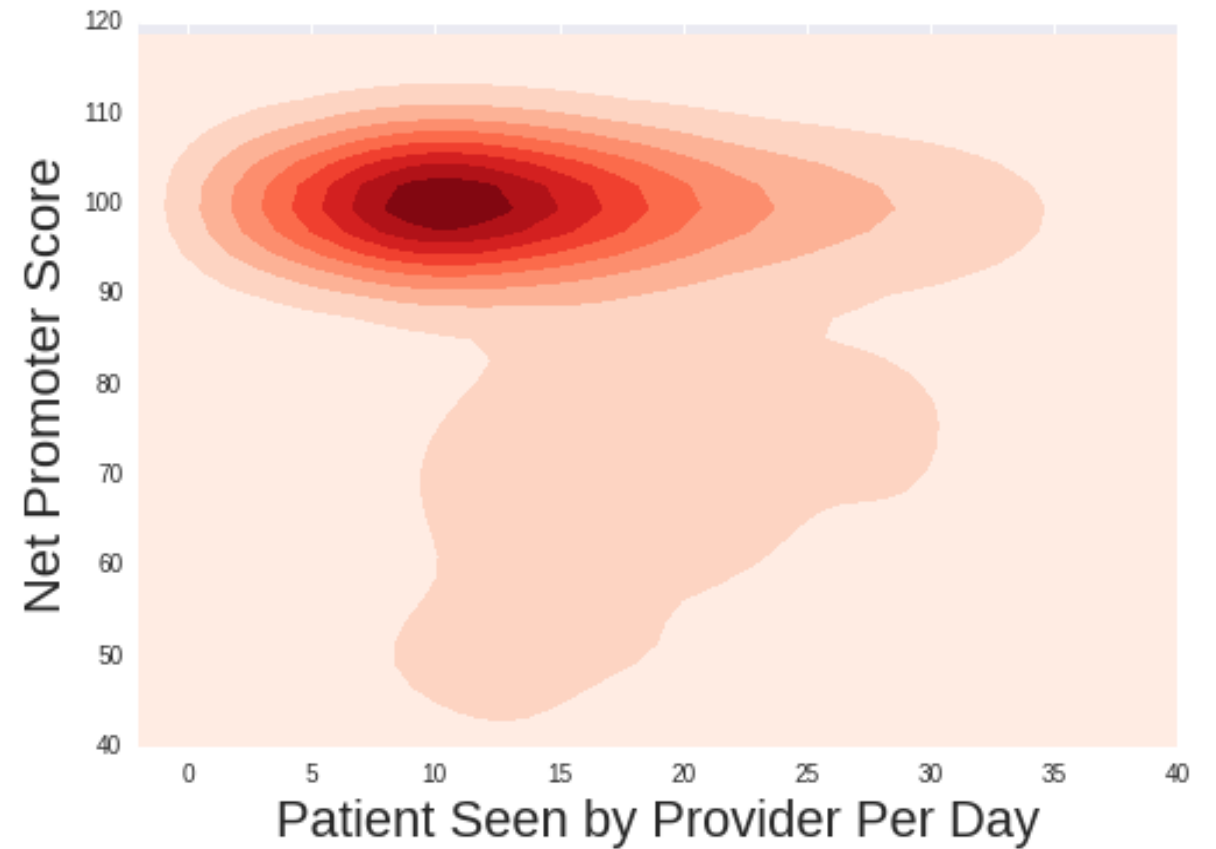


# Visualization of the Distribution

## Scatterplot Distribution

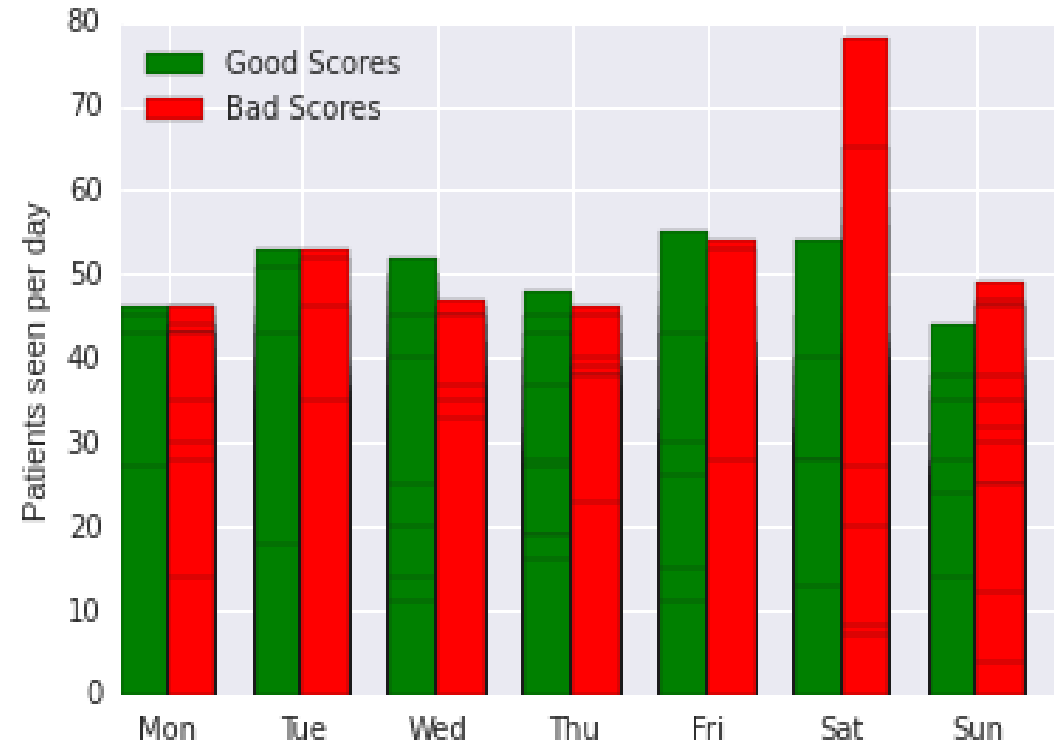


## Kernel Density Estimation

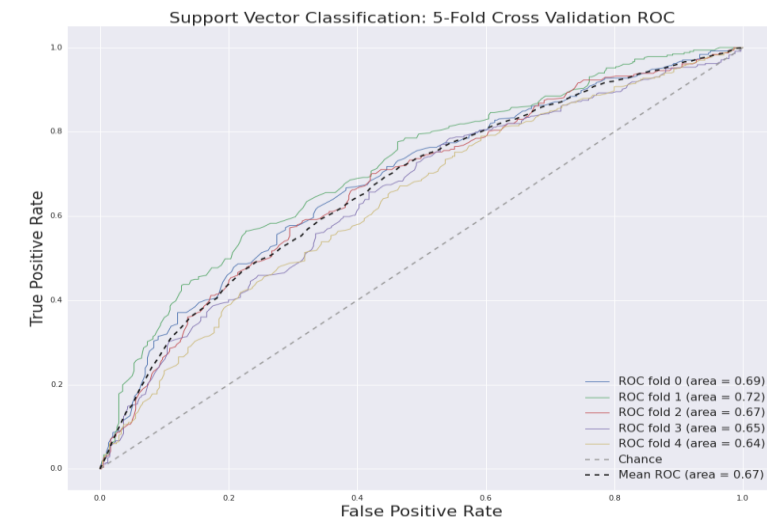
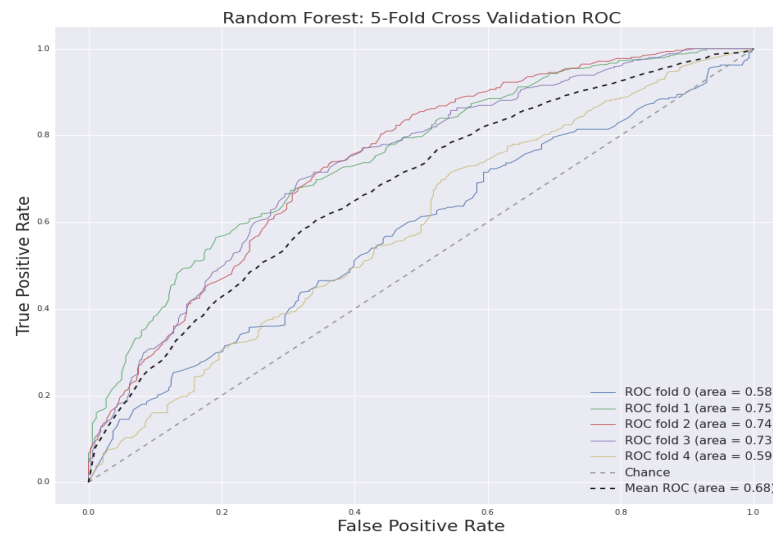
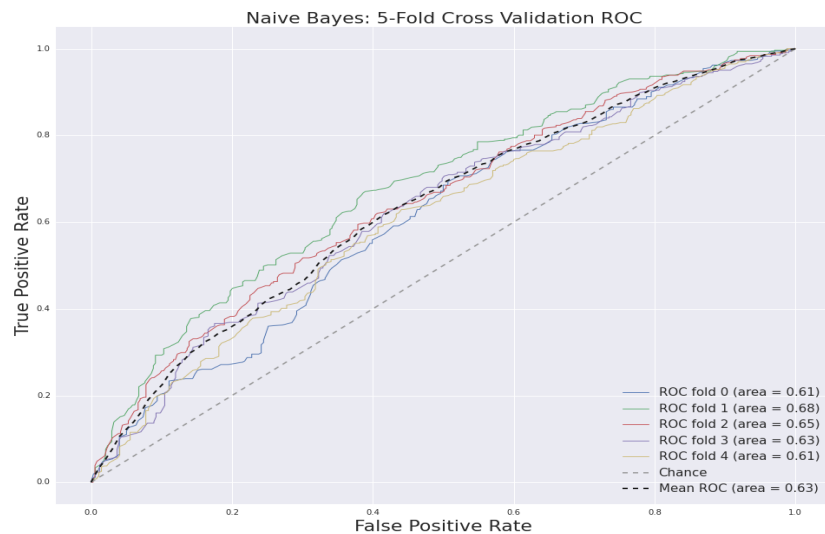


# Classifying Scores by Weekday & Patients Per Day

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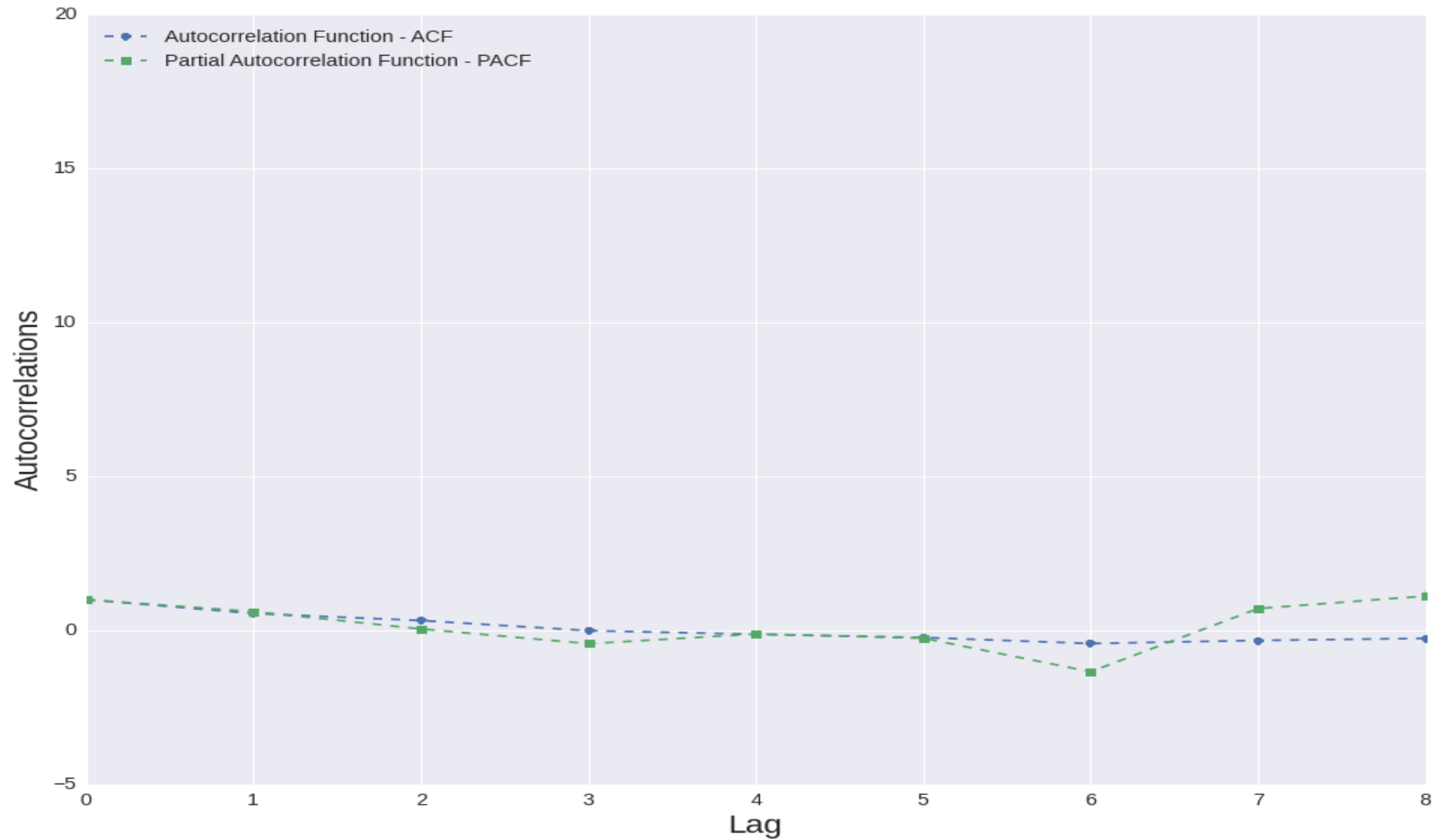


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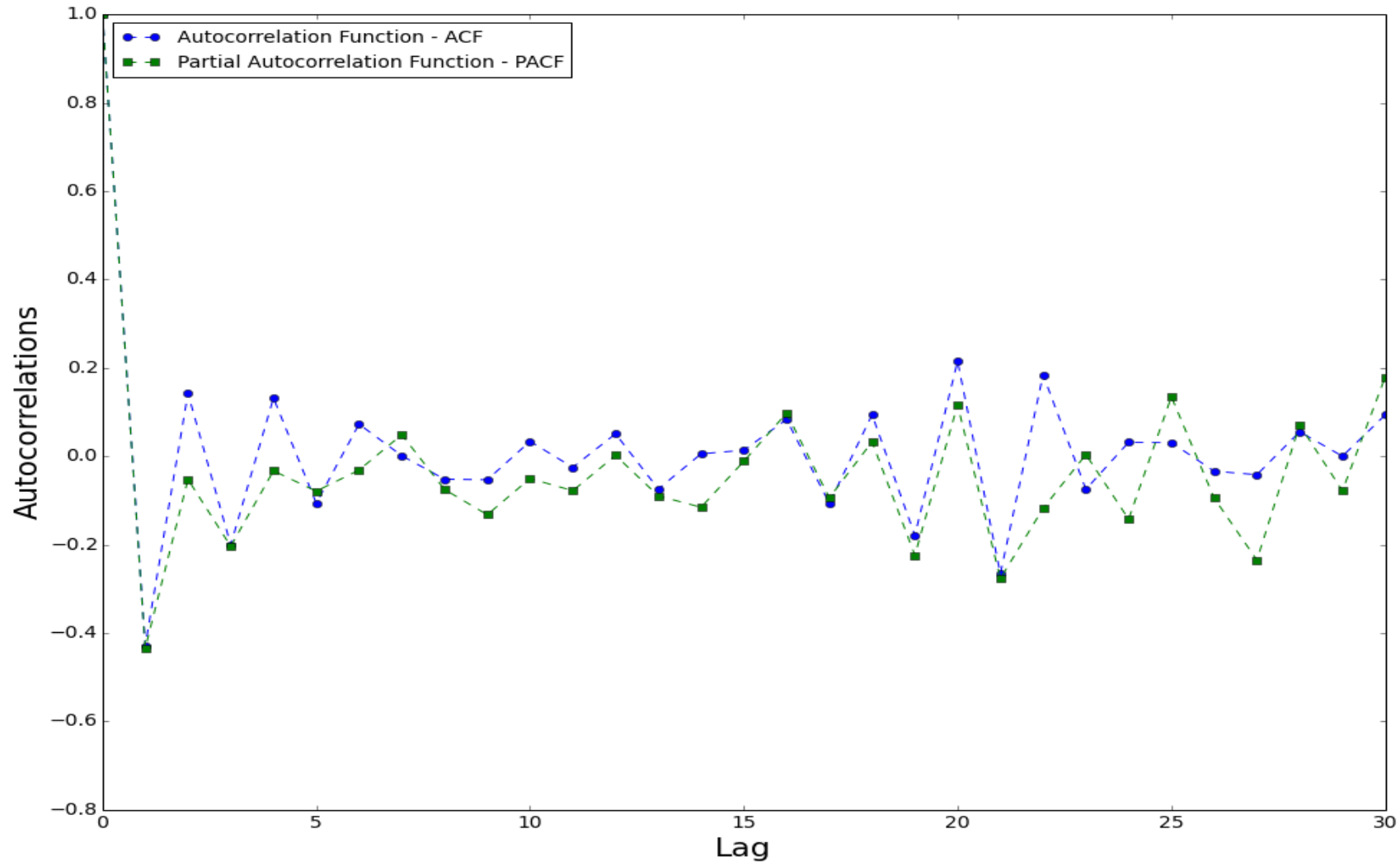


Logistic Regression Classification Summary	Precision	Recall	F1-score
Bad Score	0.60	0.45	0.52
Good Score	0.61	0.74	0.67
ROC AUC with 10-fold Cross-validation	0.68		

# Plots of ACF and PACF (Monthly)

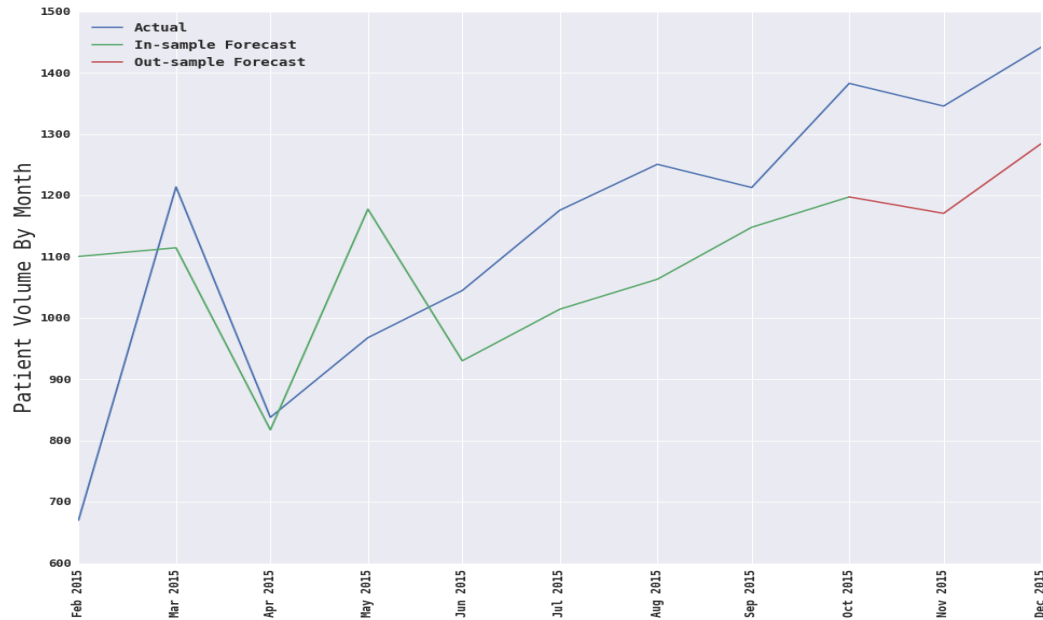


# Plots of ACF and PACF (Weekly)



# Forecasting Patient Volume (Monthly Data)

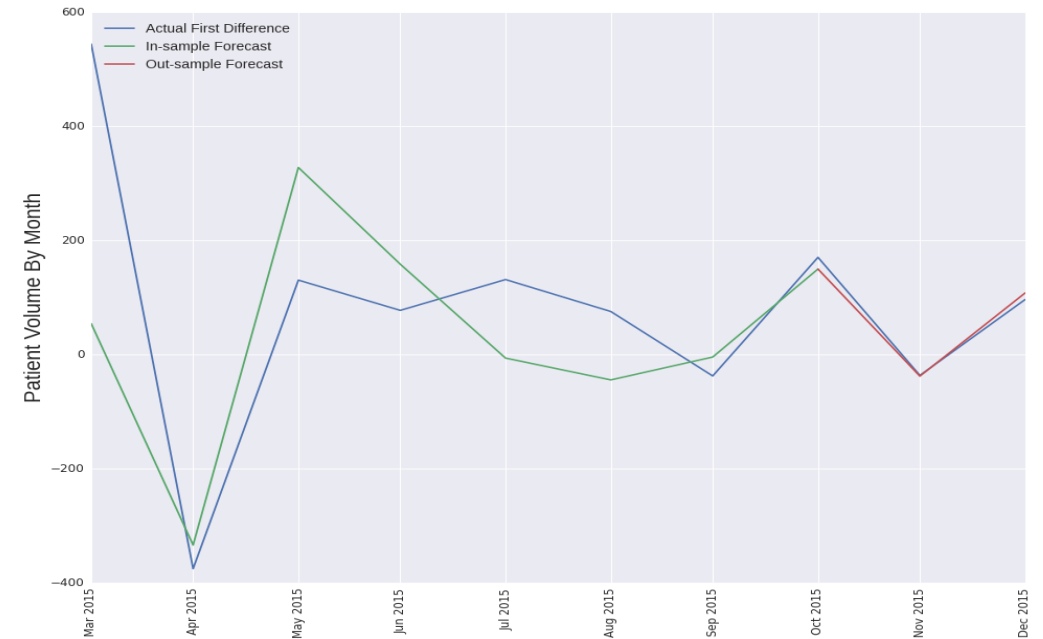
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(Non-Stationary Model)



## Approach

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**ARIMA (p=2,d=1,q=0) Model**  
(Stationary Model)

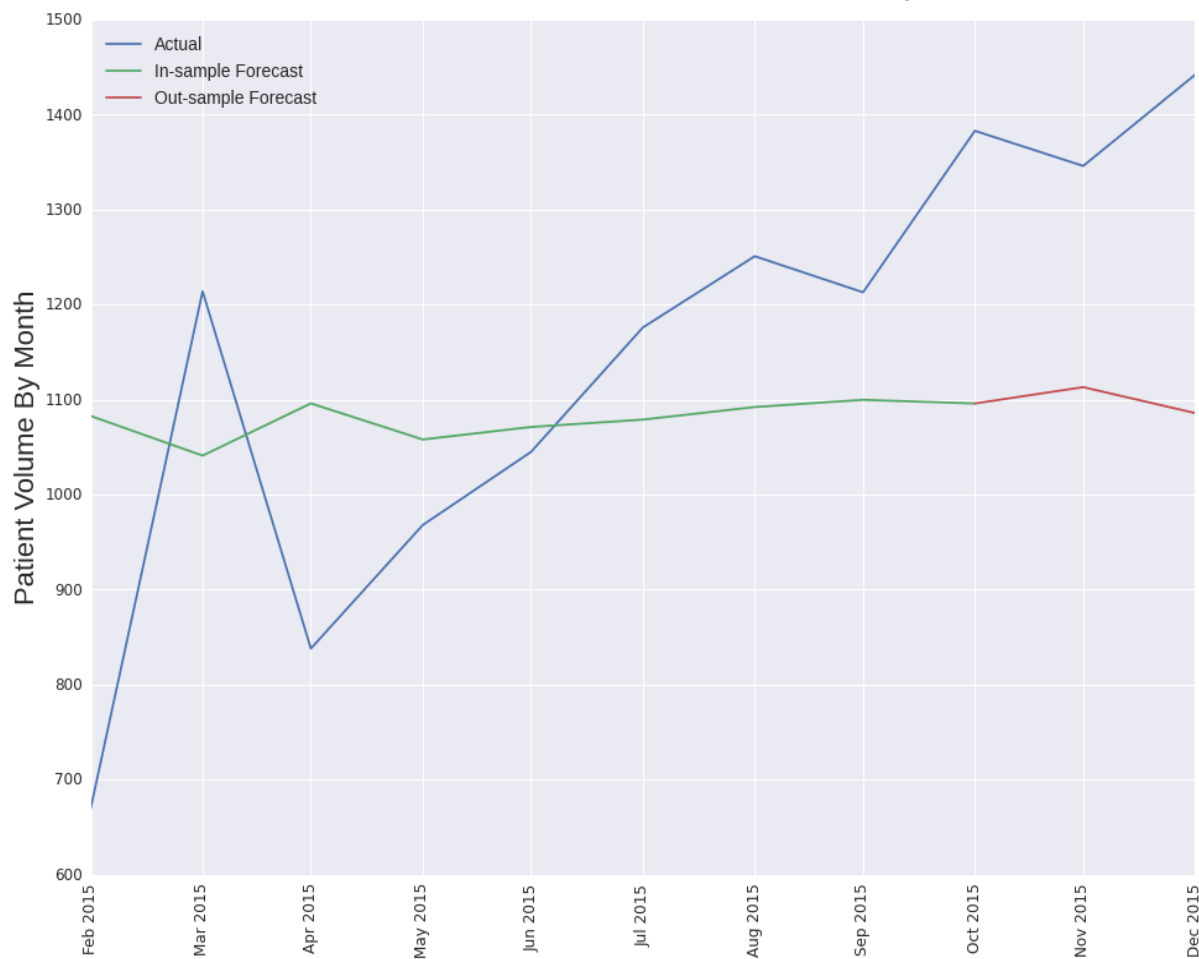


## Actionable Insights

- Mean absolute percentage error = ~15.5%
- Predicted volume outside sample (2016-01-31) is ~1487 patients

# Forecasting Patient Volume (Suboptimal Fit)

## ARIMA (1,0,0) Model for Monthly Data



## ARIMA (0,0,1) Model for Weekly Data

