# **AD 571: Business Analytics Foundations**

# LCR Session (Entire Class) Week 4 (5/30/2023)

### **AGENDA**

- Summary Weeks 1 to 3: Q & A
- Preparation for Week 4:
  - A Introduction to Lecture 07 and Lecture 08
  - (B) Individual Exercise: Tutorial: Preparation For Assignment 4 Predictive

**Tutorial: Time Series Analysis in R** 

**Tutorial: Regression Analysis in R** 

- Group Discussion Forum W4 & Quiz 4, Exercise M4: Q & A
- **3** Assignment 4: Preview & Examples (Due: 06/05/2023 at 11:59 pm ET)

### **Preparation for Week 4:**

### Module 4

May 30 - June 5

Topics: Lecture 7: Predictive Business Analytics: Basic Concepts and

Applications

Lecture 8: Model Building for Selected Predictive Business Analytics

Readings: Lecture 7:

Lecture Notes

• Evans, Chapter 8

Wickham & Grolemund, Chapter 23.3

Lecture 8:

Lecture Notes

• Evans, Chapter 9

• Wickham & Grolemund, Chapter 24.3

Tutorials: Assignment 4 Tutorial: Predictive Analytics

Regression Analysis in R

Time Series Forecasting

Discussions: Discussion 4

• Initial post due Thursday, June 1 at 11:59 PM ET.

· Respond to at least two of your classmates' posts by Monday,

June 5 at 11:59 PM ET.

Assignment 4: Predictive Analytics, due Monday, June 5 at 11:59 PM Assignments:

ET

Evans Textbook: Module 4 Exercises, due Monday, June 5 at 11:59 PM

ET

Quiz 4, available from Saturday, June 3 at 9:00 AM ET to Monday, Assessments:

June 5 at 11:59 PM ET

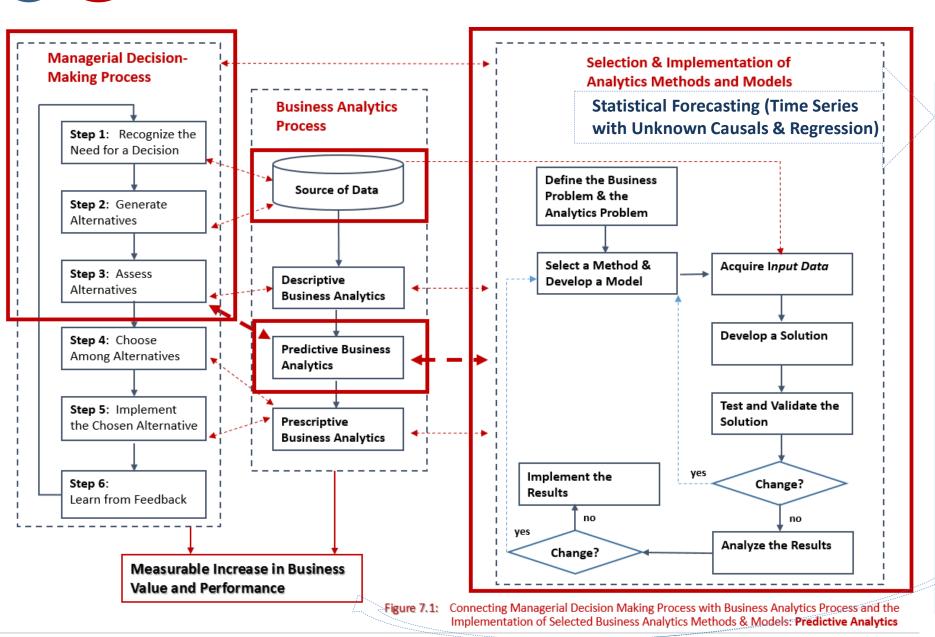
Module 4 Live Classroom Session — Tuesday, May 30, at 8:00 PM ET Live

Classroom:

Technical Session - Thursday, June 1, at 8:00 PM ET



### **Preparation for Week 4**



**Moving Averages** 

**Single Exponential Smoothing** 

**Double Exponential Smoothing** 

**Triple Exponential Smoothing** 

**Multiple Regression** 

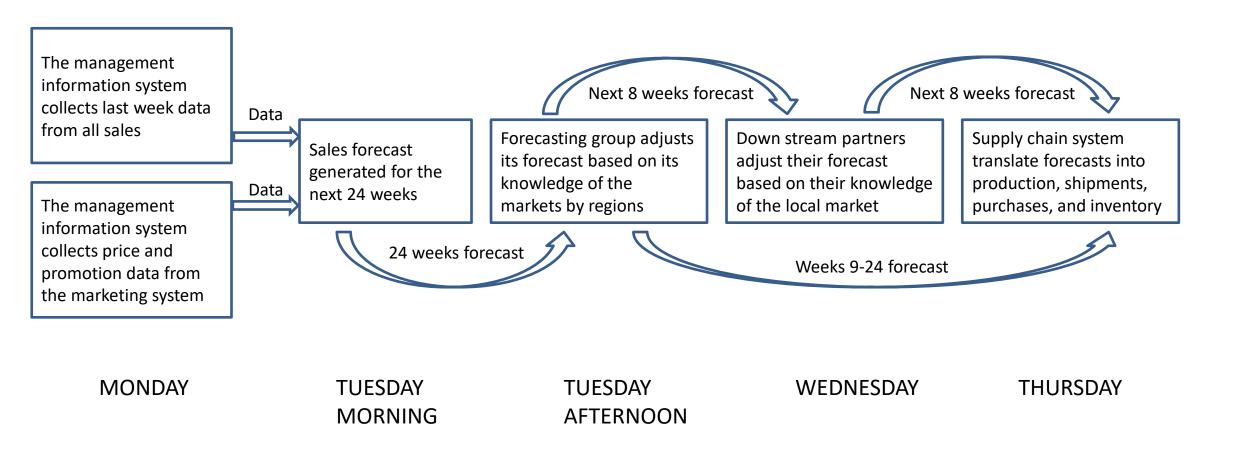
**Evaluation of the Results** 

### Recommendations:

- → Business Analytics Process(overall & next steps)
- → Managerial Decision-Making Process (overall & next steps)



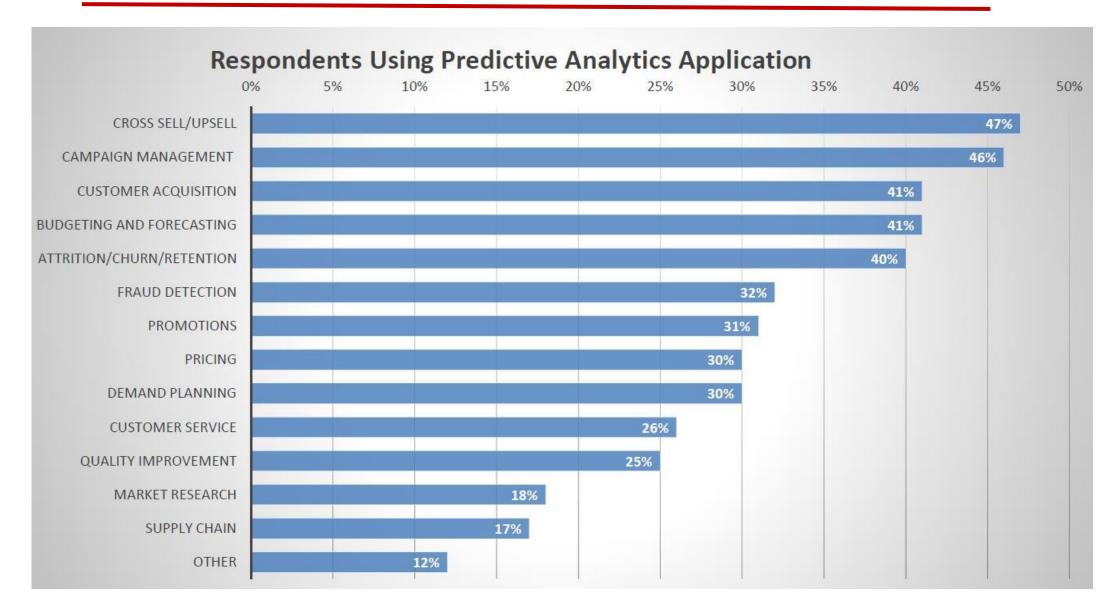
### **Preparation for Week 4**

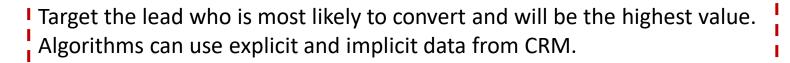






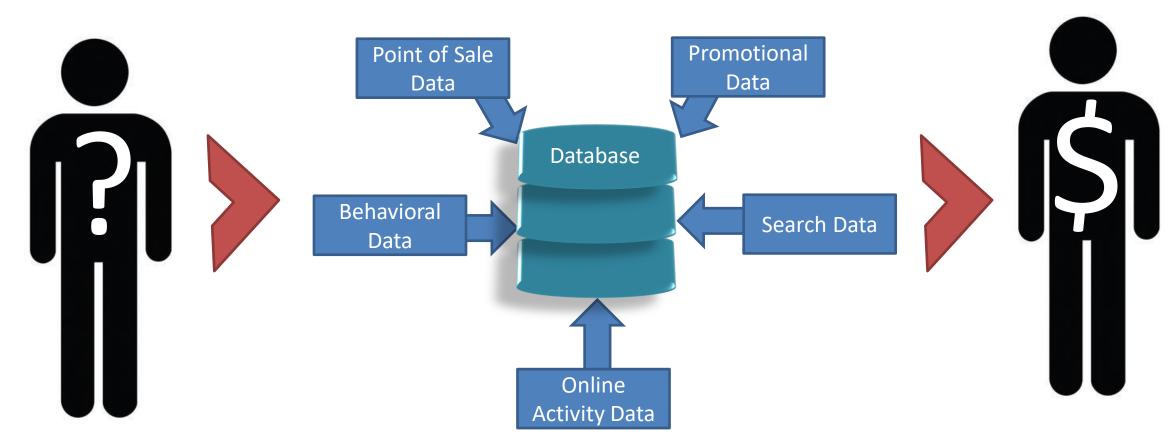
# **Preparation for Week 4: Predictive Analytics Applications**







**Problem**: Customer acquisition costs are high



We don't know customer well until data is collected



# **Predictive Analytics Applications**

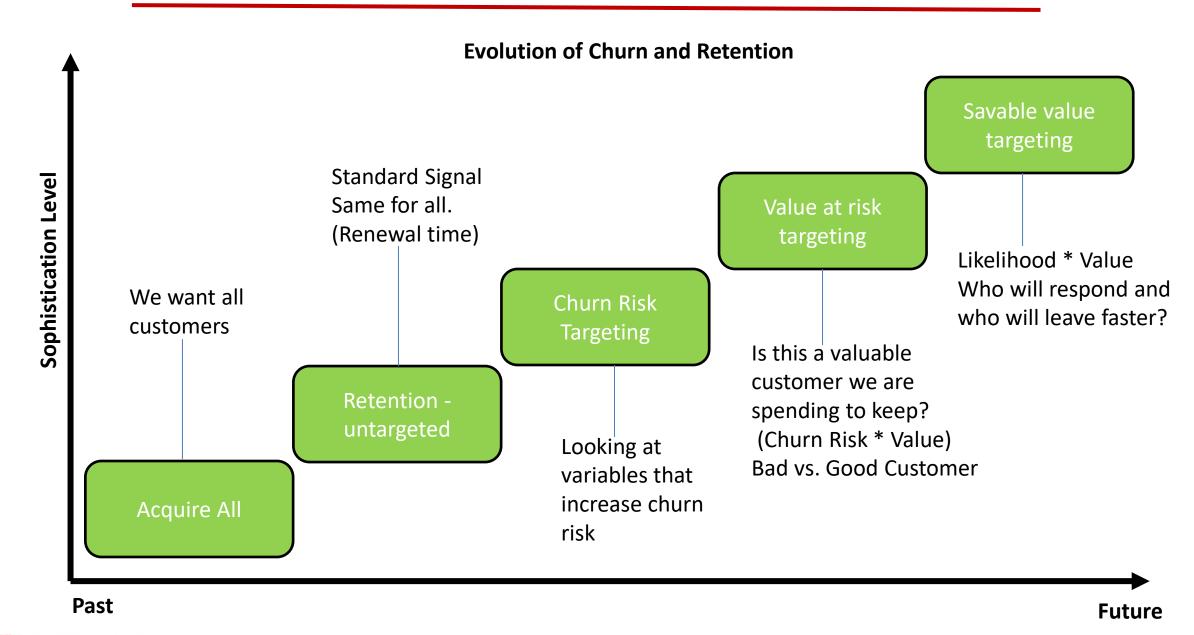
# **Attrition/Churn/Retention**







### **Preparation for Week 4:** Predictive Analytics Applications – Customer Churn





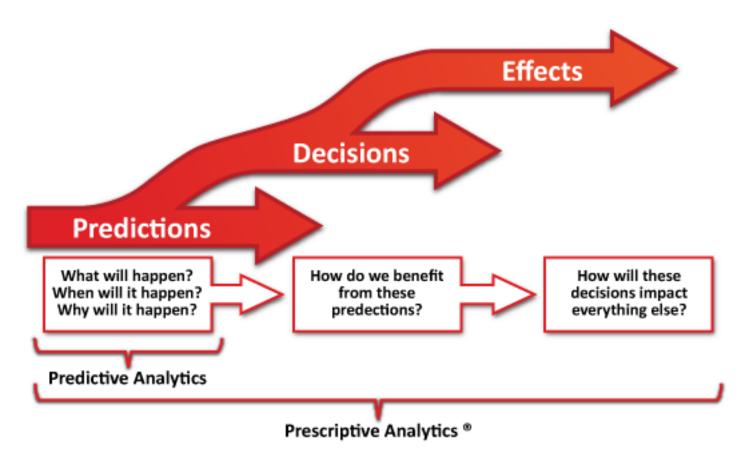


# **Preparation for Week 4: Predictive Analytics Applications**





# **Preparation for Week 4:** Predictive Analytics Applications – Risk Reduction



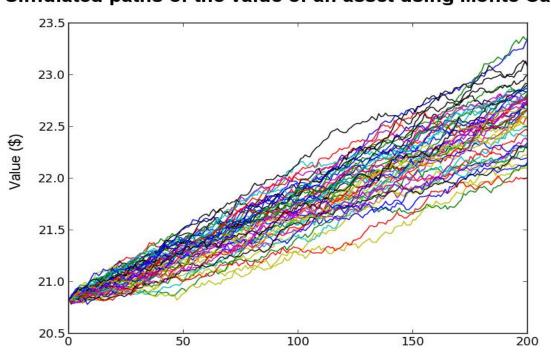
Source: https://commons.wikimedia.org/

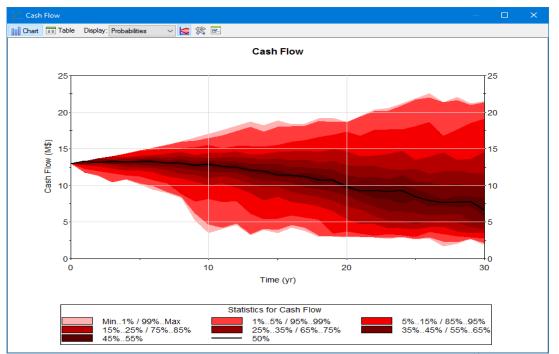




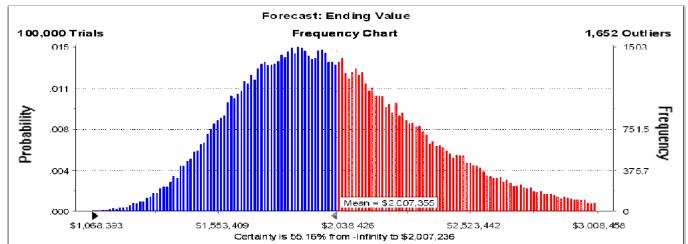
# **Preparation for Week 4:** Predictive Analytics Applications – Financial Modeling

### Simulated paths of the value of an asset using Monte Carlo





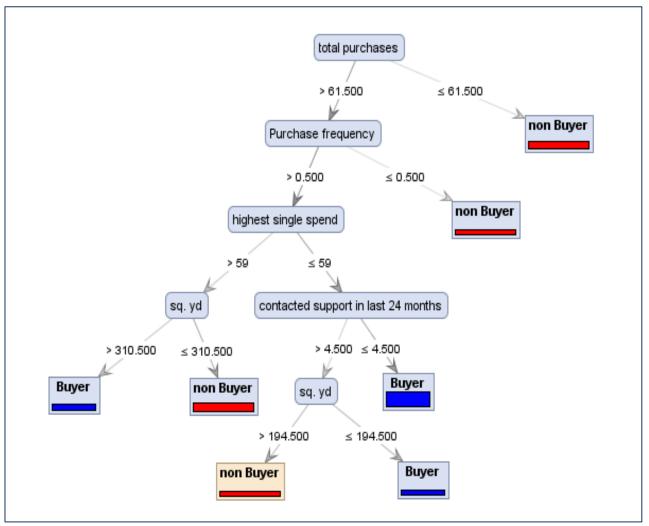
http://goldsim.com

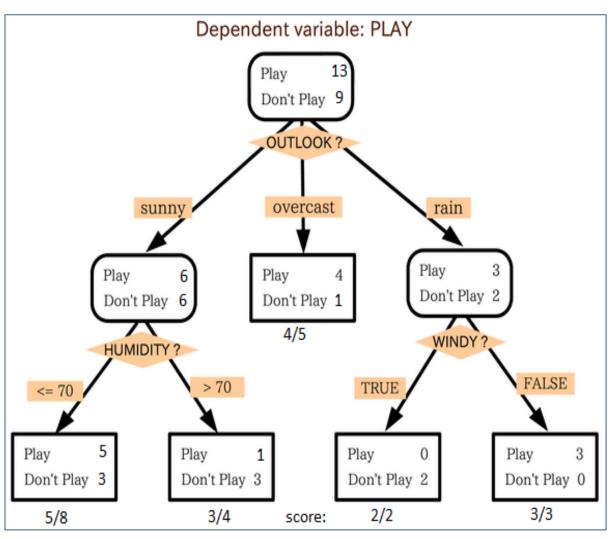






### **Preparation for Week 4:** Predictive Analytics Applications – Risk Reduction





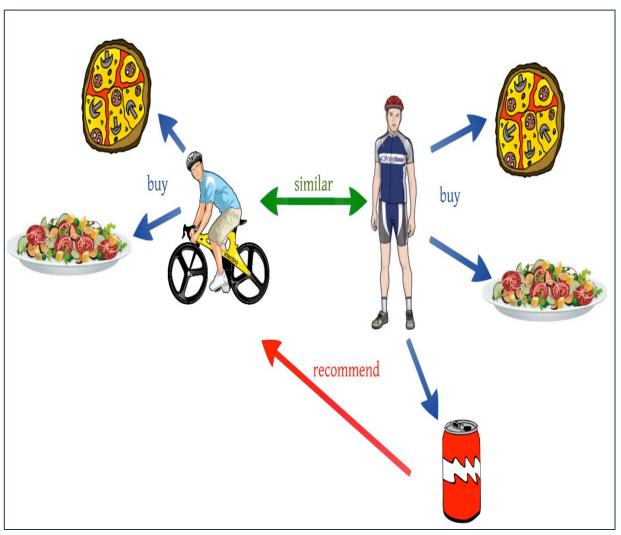
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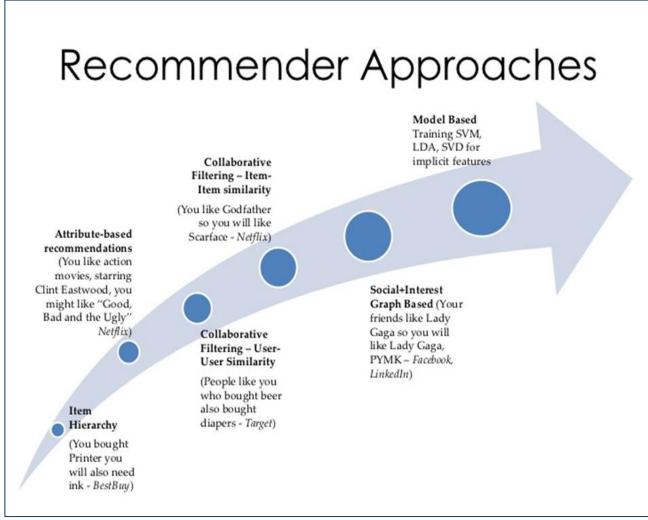
Source: http://stackexchange.com





# **Preparation for Week 4:** Predictive Analytics Applications – Recommendations





Source: http://medium.com

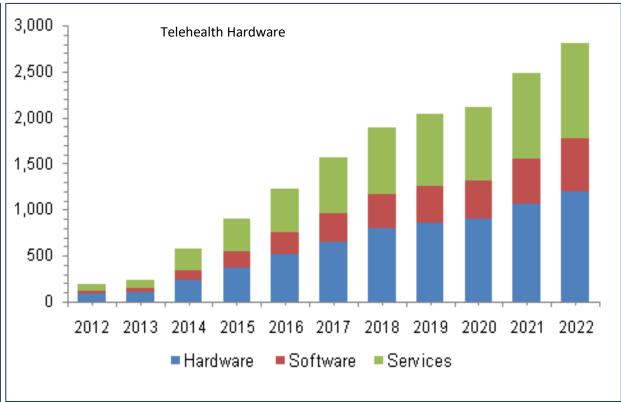
Source: http://Sflscientific.com





### **Preparation for Week 4:** Predictive Analytics Applications – Market Analysis





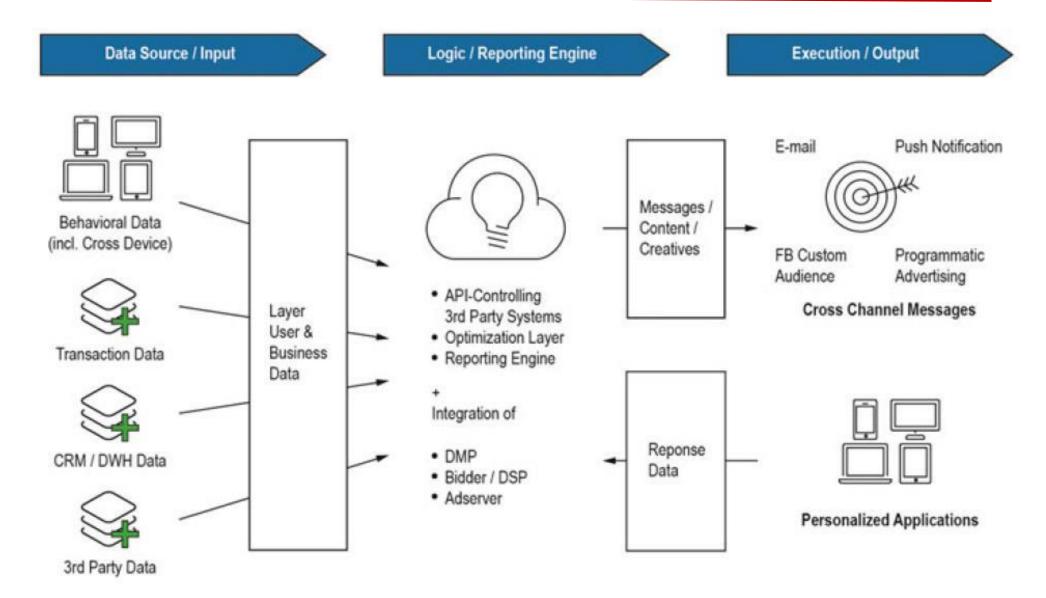
Source: http://www.gminsights.com

Source: http://chironhealth.com





### **Preparation for Week 4:** Predictive Analytics Applications – Digital Advertising



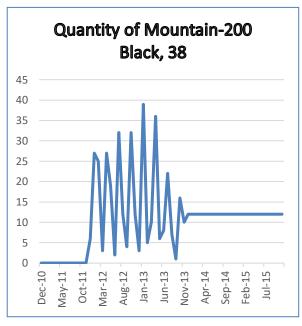
Source: Heinemann, F.(2016) Driving Performance with Programmatic CRM



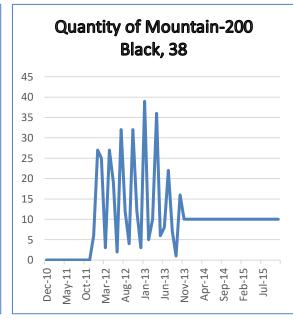


### **Preparation for Week 4: Time Series**

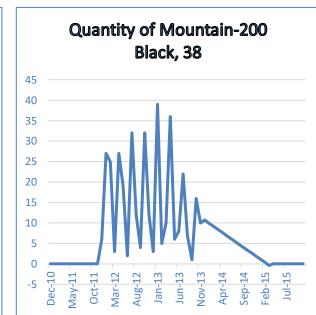
### Single Exponential Smoothing



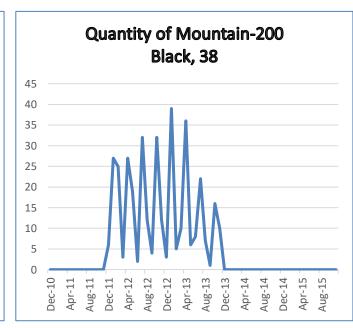
**Moving Averages Smoothing** 



# Double Exponential Smoothing



# Triple Exponential Smoothing



182,425
9.04737
144.917
6.62427
59,9322
1.87288

Mean Absolute Percentage Error (MAPE)	223.5788
Mean Absolute Deviation (MAD)	12.625
Mean Square Error (MSE)	290.5625
Tracking Signal Error (TSE)	0.792079
Cumulative Forecast Error (CFE)	10
Mean Forecast Error (MFE)	0.3125

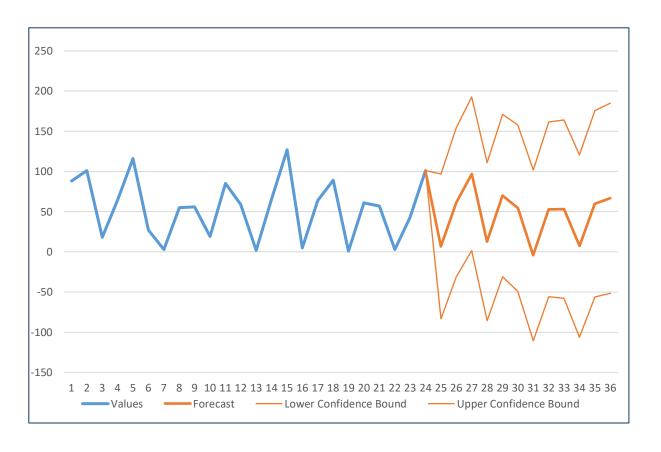
234.7263
9.811871
140.8039
-6.78919
-66.6146
-2.01862

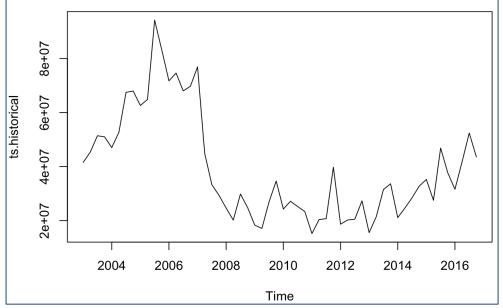
Mean Absolute Percentage Error (MAPE)	100
Mean Absolute Deviation (MAD)	11.0303
Mean Square Error (MSE)	266.2424
Tracking Signal Error (TSE)	33
Cumulative Forecast Error (CFE)	364
Mean Forecast Error (MFE)	11.0303

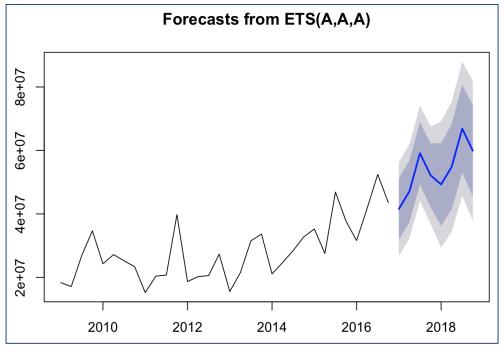




# **Preparation for Week 4: Tutorial For Predictive Analytics: Time Series Visualization**











### **Preparation for Week 4:**

### **Predictive Analytics Applications – Multiple Regression**

Min 1Q Me	edian	3Q	Max			
-20614597 -1849007	0	1740805	14013527			
Coefficients:						
		Estimate			Pr(> t )	
(Intercept)	55087	544.17370	59902286, 93867	0.920	0.36053	
BuildingClassFinalRollC0	-5118	829.05153	5291132.55548	-0.097	0.92318	
BuildingClassFinalRollC1	19729	963.67852	5201342.63481	0.379	0.70546	
BuildingClassFinalRollC2	-915	817.86150	5416380.66169	-0.169	0.86616	
BuildingClassFinalRollC3	15683	124.41688	5343466.16755	0.293	0.76993	
BuildingClassFinalRollC4	7878	887.78238	5025496.19446	0.157	0.87581	
BuildingClassFinalRollC6	-53696	546.61639	6332651.59816	-0.848	0.39901	
BuildingClassFinalRollC7	18073	141.72312	5225211.73749	0.346	0.73036	
BuildingClassFinalRollD1	48886	584.81144	6252896.69410	0.782	0.43662	
BuildingClassFinalRollD3	66668	868.66283	7723701.02615	0.863	0.39062	
BuildingClassFinalRollD6	-21913	591.13592	6607430.12443	-3.317	0.00137	**
BuildingClassFinalRollD7		822.25663	5797734.15671	0.416	0.67878	
BuildingClassFinalRollD9	8022	386.72043	6343883.57492	1.265	0.20969	
BuildingClassFinalRollR9	102554	115.14753	8060905.28646	1.272	0.20697	
BuildingClassFinalRollRR	118994	177.69424	7855328.16592	1.515	0.13376	
ResidentialUnits	-248	544.32987	98603.65269	-2.521	0.01370	se
CommercialUnits	2760	046.95689	646323.23727	0.427	0.67045	
GrossSqFt	X1257775	794.17103	95.95051	8.277	0.00000000000232	岩岩岩
SaleDate		0.03094	0.01027	3.012	0.00348	ste ste
YearBuilt	-508	845.34076	31243.58536	-1.627	0.10759	
Signif. codes: 0 '***' (	0.001	**' 0.01	'*' 0.05'.' 0.1	''1		
Residual standard error:	481900	on 80 d	earces of freedo			
Multiple R-squared: 0.84			R-squared: 0.8	079		
F-statistic: 22.91 on 19					222	

### Why we will use Adjusted R-squared

- R<sub>2</sub> will often increase if you add more predictors to the model
- A large R-squared may be an outcome of adding more predictor to the model
- We make use of the adjusted R-squared to account for the number of predictor in the model

### How do I know which variables to add?

Generally, additional variables should have:

- Theoretical validity
- Explanatory power
- Avoid high correlation with another explanatory variable
- >.6 correlation is the standard cut-off.

# **Preparation for Week 4:** Group Discussion Forum W4

### Please select just one of the bullet points listed below:

- Discuss a use case for a regression model that can be applied to your professional field
- Discuss a use case for smoothing techniques that can be applied to your professional field
- Discuss the significance of the different types of regression models
- Discuss the significance of the different types of smoothing techniques
- Discuss topics of your choice from Lecture 7 and Lecture 8
- Discuss topics of your choice from Evans Chapters 8 and 9

# **Preparation for Week 4:** Quiz 4: Q & A

Quiz 4 consists of 10 multiple choice and true-false questions.

Quiz 4 will cover information from the

- Online lecture notes (Lecture 7 and Lecture 8),
- Online discussion Week 4.

The quiz is open-book.

### **Format**

- You will have 30 minutes to complete the quiz. There is a clock in the upper right corner of the screen keeping time for the exam.
- You can take the quiz only once.
- > Each question will be delivered one at a time.
- You can revisit the questions and change your answers as many times as you want before submitting the exam.

# **Preparation for Week 4:** Module 4 Exercise: Q & A

# Questions are located at the end of Chapter 8 Section: Problems and Exercises

Group	1 - Roman	Grou	p 2 - Vivek
Due Date	Monday, June 5, 2023	Due Date	Monday, June 5, 2023
Student	Chapter 8	Student	Chapter 8
Yurui Chen	7	Audrey Chan	7
Victor Brice Fedjo Yemele	8	Flori-Ann DeLa Cruz	8
Jonathan Garrison	9	Marcel Fernandes Silva	9
Payton Hatcher	10	Avirul Islam	10
Faria Hossain	2	Jiarui Lin	2
Jiamin Li	3	Joyce Machau	3
Kaitlynn Nguyen	4	Nicole Matarazzo	4
Miranda Petrillo	5	Timothy Olakunle	5
Haoqiang Qi	6	Olu Olayeye	6
Nesteshia Riddell-Dell	20	Sandhya Ramani	20
Cassandra Simoneau	21	Sri Amruta Sripada	21
Yang Yang	22	Samuel Stevens	22
Samuka Yekeh	23	Jack Swartz	23

### **Business Running Case—Targeted Outcomes:**

### **Application of Predictive Analytics**



W-2

**Term Project A1**: Data Import and Visualization

**Term Project A2**. Data Manipulation In R

W-3

**Term Project A3**. Descriptive Analytics Techniques In R

W-4

**Term Project A4**. Predictive Analytics Techniques In R

W-6

**Term Project A5**. Prescriptive Analytics Techniques in R and Excel

W-7

Term Project A6. Term Project Final Presentation Model Deployment

### V-Lab Instructions

All course tools and SQL Server can be accessed from the V-Lab, which is included in tuition.

Assignment 4 Requires access to R Studio and SQL Server

### **Assignment 4: Predictive Analytics**

**Assignment 4 Objective:** Prepare a managerial report, starting with an executive summary; expected length up to 4 pages APA format, excluding cover page, table of content, and appendixes.

- 1. Perform time series analysis on the total dollar amount of residential real estate sales on your neighborhood. Use sales beginning in the year 2009 to develop your model. Develop a forecast for the next 8 quarters of sales.
- 2. Use a multiple regression model to come up with another forecast for the next 8 quarters of sales. Include time and seasonality. Use sales beginning in the year 2009 to develop your model.
- 3. Use a multiple regression model to determine the sale of a given residential property in your neighborhood. Include:
  - a. Sale Date
  - b. Year built
  - c. Building type (categorical)
  - d. Gross Square Feet
  - e. Number of Units
- 4. According to your model from (3), what are the most and least useful predictors of the amount of a sale?
- 5. Are there any redundant independent variables in your model from (3)? How can you tell?
- 6. According to your model from (3), which properties were the biggest bargains and which were the most overpriced? How might you account for these disparities?
- 7. Write 3-4 pages summarizing your findings with a focus on the output, interpretation of the output, and what the insights mean for our decision-making process

[Hint: your model may be more accurate if you discard data from prior to the housing market crash]

LOCATION OF THE INFORMATION AND SOFTWARE

#### **Course Website**

Page on Blackboard: "Assignments">>>Assignment 4

### **Tutorials Section**

Link

Tutorial – Assignment 4

Q & A