













Inspire...Educate...Transform.

Data Science: Big Picture

Also, introduction to CPEE program

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May 31, 2015

THE OUTPUT



Multiple forms

Rules

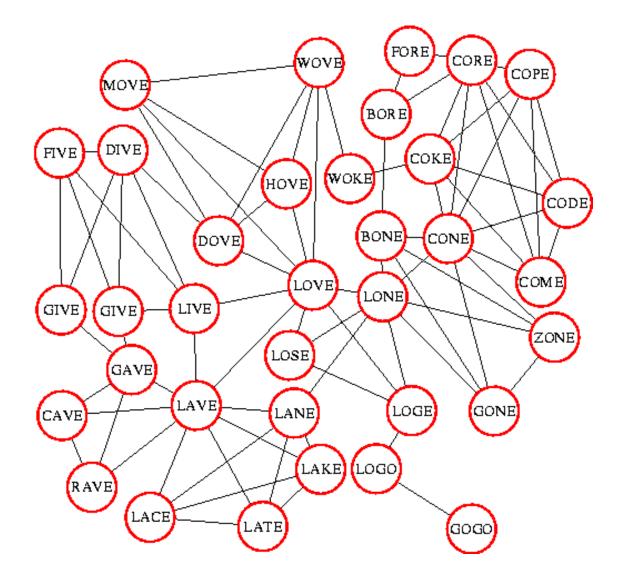
-If x1, x2, x3 then Y

Equations

$$-Y=f(x)$$



Graphs





Output

• Similarities

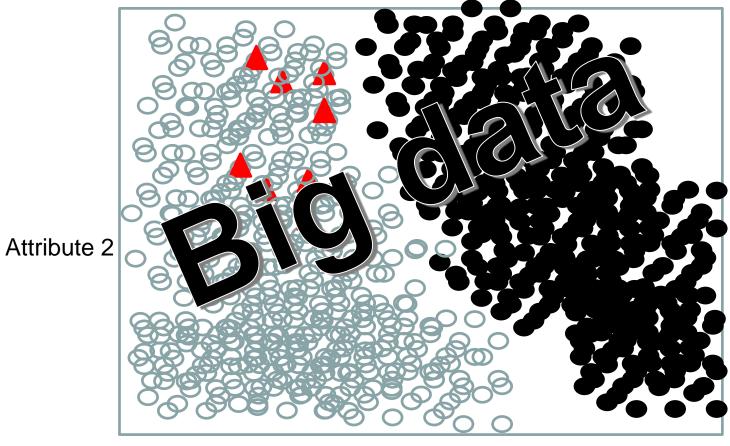
Blackbox



TYPES OF CHALLENGES







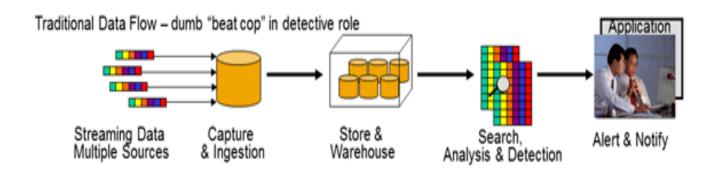
Attribute 1

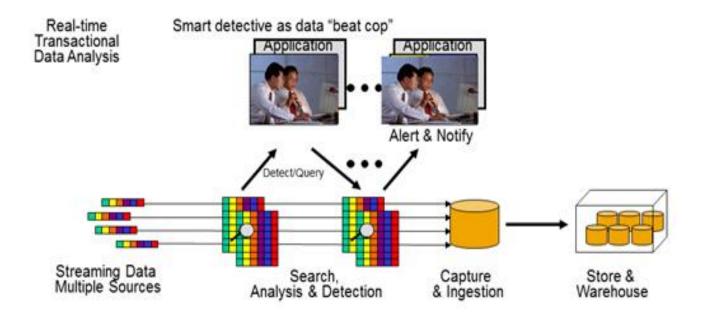


Velocity

• Every minute, 100 hours of video on YouTube, 200 million emails and 300,000 tweets.

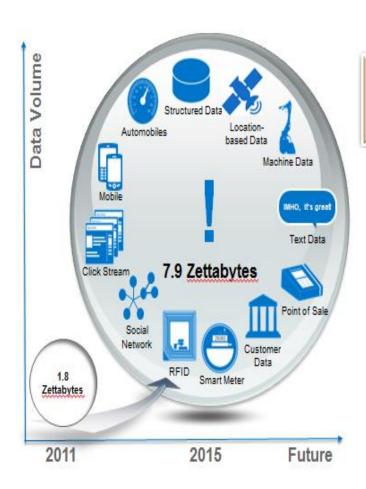






9

Variety



- ➤ 1 Terabyte = 1024 Gigabytes
- > 1 Petabyte = 1024 Terabytes
- > 1 Exabyte = 1024 Petabytes
- > 1 Zettabyte = 1024 Exabytes

Variability

- Understanding unstructured information is complex
 - My in-laws are as sweet as Nazis
 - Police is tracking terrorists with bombs
 - He eats, shoots and leaves
 - Great
 - "Delicious muesli from the @imaginarycafe- what a great way to start the day!
 - "Greatly disappointed that my local Imaginary Cafe have stopped stocking BLTs."
 - "Had to wait in line for 45 minutes at the Imaginary Cafe today. Great, well there's my lunchbreak gone..."



Veracity

• Lots of cleaning and noise removal is needed.



Value

• Healthcare related big data efforts "could account for \$300 billion to \$450 billion in reduced health-care spending

• Data on its own is worthless. The value lies in rigorous analysis of accurate data for valuable insights



Visualization

• Do visualizations really make a big difference?



How many numbers are less than 10

```
83 11 70 27 66 67 12 96 48 70 97 1 64 28 94 51 46 52 90 82 92 16 3 98 62 21 7 68 11 71 96 79 27 22 3 47 59 94 48 11 11 54 8 51 17 9 96 15 7 11 58 52 86 68 60 73 20 15 4 19 3 78 82 9 54 60 75 88 42 88 49 65 44 65 44 25 14 26 17 81 48 93 10 88 67 87 11 34 35 55 74 17 11 25 39 96 26 39 88 59
```

You have 8 seconds



How many numbers are less than 10

```
83 11 70 27 66 67 12 96 48 70 97 1 64 28 94 51 46 52 90 82 92 16 3 98 62 21 7 68 11 71 96 79 27 22 3 47 59 94 48 11 11 54 8 51 17 9 96 15 7 11 58 52 86 68 60 73 20 15 4 19 3 78 82 9 54 60 75 88 42 88 49 65 44 65 44 25 14 26 17 81 48 93 10 88 67 87 11 34 35 55 74 17 11 25 39 96 26 39 88 59
```

You have 4 seconds



I]	II		III		ſΨ	
X	Y	Х	Y	X	Y	X	Υ	
10.0	8.04	10.0	9.14	10.0	7.46	8.0	6.58	
8.0	6.95	8.0	8.14	8.0	6.77	8.0	5.76	
13.0	7.58	13.0	8.74	13.0	12.74	8.0	7.71	
9.0	8.81	9.0	8.77	9.0	7.11	8.0	8.84	
11.0	8.33	11.0	9.26	11.0	7.81	8.0	8.47	
14.0	9.96	14.0	8.10	14.0	8.84	8.0	7.04	
6.0	7.24	6.0	6.13	6.0	6.08	8.0	5.25	
4.0	4.26	4.0	3.10	4.0	5.39	19.0	12.50	
12.0	10.84	12.0	9.13	12.0	8.15	8.0	5.56	
7.0	4.82	7.0	7.26	7.0	6.42	8.0	7.91	
5.0	5.68	5.0	4.74	5.0	5.73	8.0	6.89	

- Mean of x: [9999]
- Variance of x: [11 11 11 11]
- Mean of y: [7.5 7.5 7.5 7.5]
- Variance of y: [4.127 4.128 4.123 4.123]
- Correlation of x-y: 0.816 0.816
 0.816 0.817



• Equation of regression line for

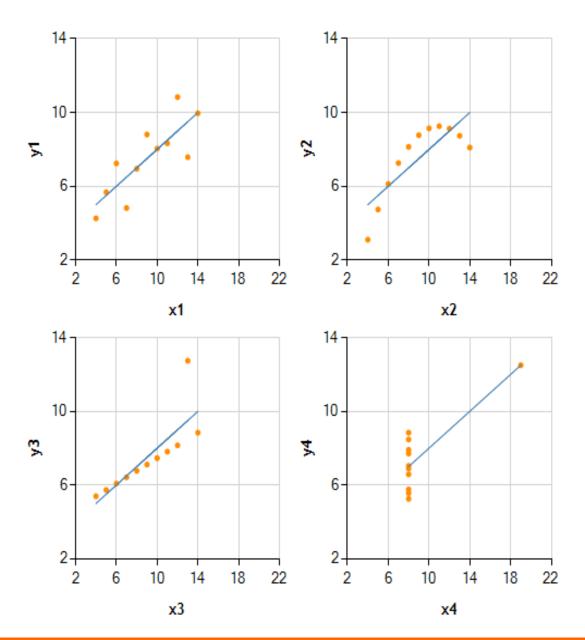
$$-1$$
: Y = 3 + 0.5X; r2: 0.67

$$-2$$
: Y = 3 + 0.5X; r2: 0.67

$$-3$$
: Y = 3 + 0.5X; r2: 0.67

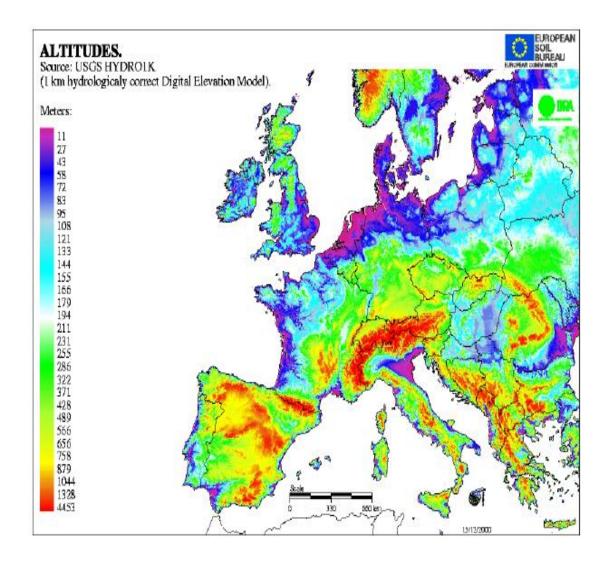
$$-4$$
: Y = 3 + 0.5X; r2: 0.67







Some bad visualizations





Great visualization An example



22



Angola





Columbia





Europe





China





Burkina_Faso





Brazil





USA



7 Vs.

- Volume
- Velocity
- Variety
- Variability

- Veracity
- Visualization
- Value



SOLUTION ARCHITECTURE



ROI understanding

- Business problem
- Current approach
- Advantages of data way of thinking and what problem you are solving
 - -Fraud detection



Feature engineering

• Can I add, transform existing attributes to generate new attributes



Getting the data into a structured form

Known/Easy to measure	Known/Easy to measure	Known/Easy to measure	Difficult to measure

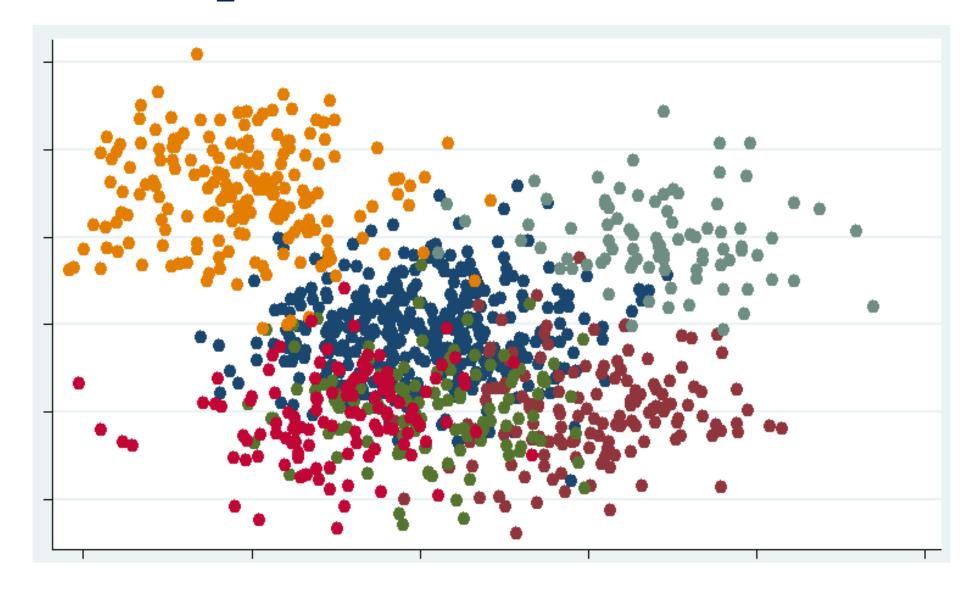


Sharpening the data

- Missing values
- Type conversions (based on domain)
- Retaining important attributes



Data exploration





Model building





Story telling

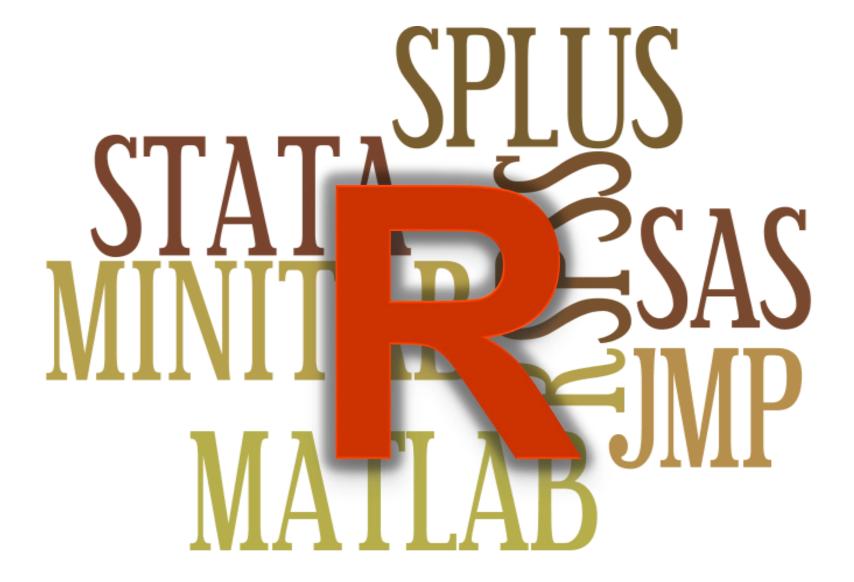




ECO SYSTEM

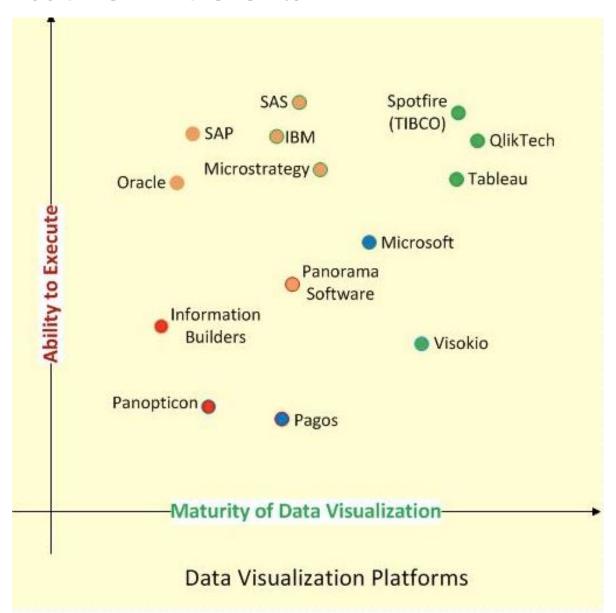


Data Science Environment



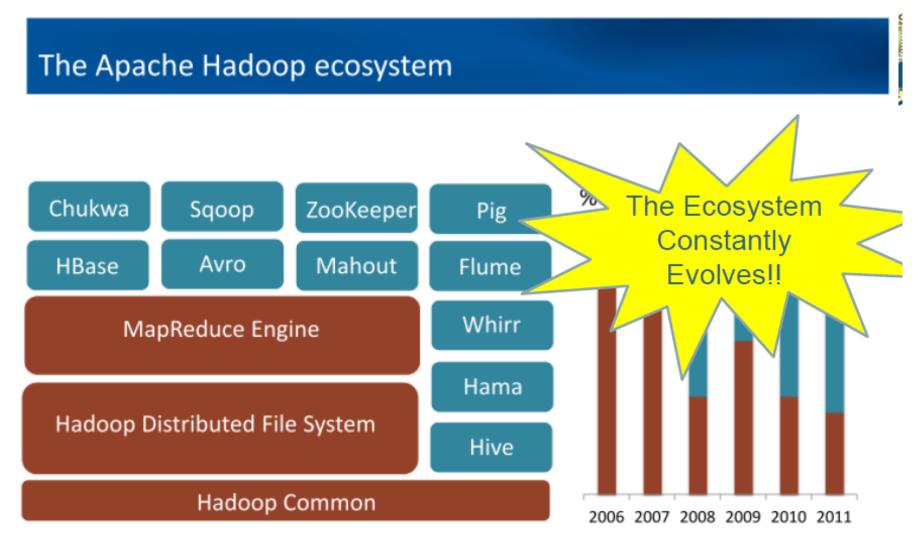


Visualization tools





What to implement







Learn enough domain

- Learn Enough Domain and look accessible in the meeting
 - -http://www.ibm.com/analytics/us/en/sol utions/index.html



Define the problem

- Goal
- Assumptions
- Process
- Business use



A definition

• We will identify customers who are likely to buy in the next campaign (or)



A better definition

 We shall identify most likely target customers for a new campaign based on similar campaigns of the past. We assume that demographic and sales habits define behavior towards the campaign. We will use demographic and sales data of the past campaigns to unearth relationship between the known characteristics of a customer to her reaction to a specific campaign.



Business use

- The customer can reduce the number of contacts while not compromising on revenues
- There are more complex cases where the business use is not obvious.
 - Telecom use
 - Clinical trials



ERROR METRICS



A rare disease

• 1 in 100,000 get it

- The model is "No body has it"
 - –What is the accuracy?
 - 99.999



Quality of the analysis

Types of errors in classification

	Predicted positive	Predicted negative
Actual positive	TP: 500	FN: 400
Actual negative	FP: 100	TN: 9000



Error metrics: Classification

• Accuracy is the percent of the predictions that were correct?

-The "accuracy" is (9,000+500) out of 10,000 = 95%



More on error measures

• Sensitivity, true positive rate, or the recall rate measures the proportion of actual positives which are correctly identified as such

$$Recall = Sensitivity = P(\hat{Y} = 1|Y = 1)$$



Precision

 Precision is how many of my predicted positives are actually positives

• Precision = P(Y=1| =1)



• In some business cases, both precision and recall may be important. Then people use F1 statistic defined by

• F1 Statistic = Metric =
$$\frac{2PR}{P+R}$$



Specificity

• Measures the proportion of negatives which are correctly identified as such (e.g. the percentage of healthy people who are correctly identified as not having the condition, sometimes called the *true negative rate*).

• Specificity=
$$P(Y=0|Y=0)$$



Summary

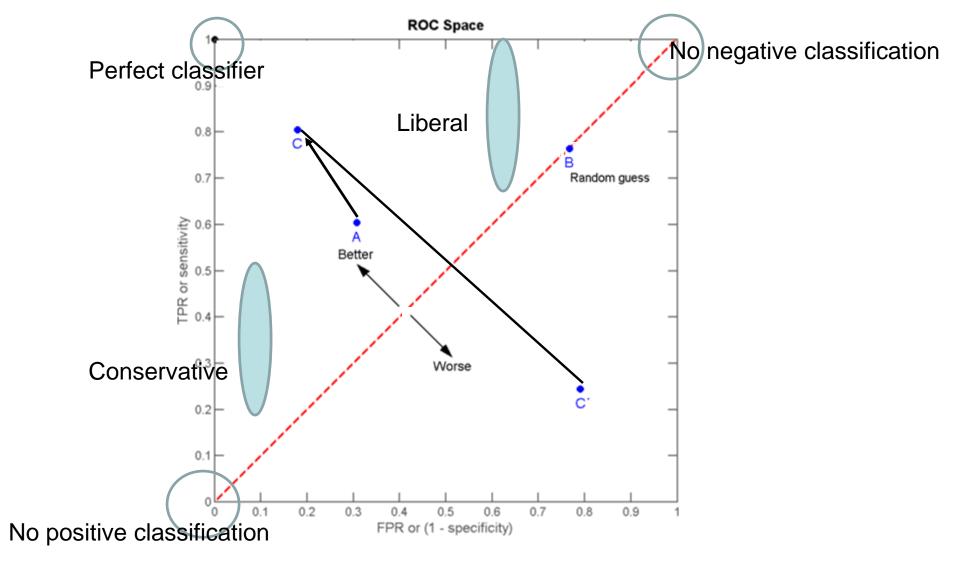
	Р		
A	63	37	
	28	72	

TPR=Sensitivity=Recall=0.63
FPR=1-specifificty=0.28
Precision= 0.69
F1=0.66
Accuracy= 0.68

False positive rate = Percentage of negatives incorrectly classified



Receiver operating characteristic curve





Reference

https://cours.etsmtl.ca/sys828/REFS/A1/Fawcett_PRL2006.pdf



Lift curves: Another goodness metric

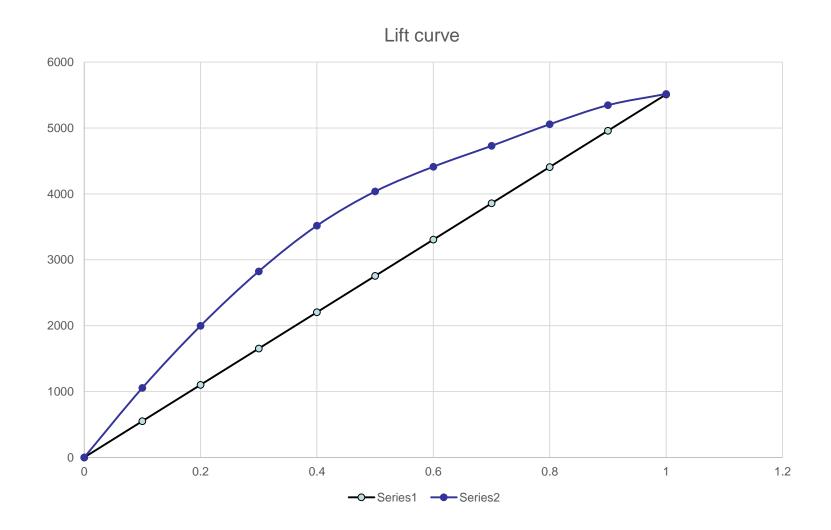
- An analysis is done to identify potential customers. Around 47% of potentials are customers.
- Later it is applied on real data.
- The potentials are sorted in the order of the probability of becoming a customer based on the model



Bins	Random	Model
1162	551	1056
1162	551	942
1162	551	826
1162	551	694
1162	551	519
1162	551	375
1162	551	317
1162	551	328
1162	551	289
1162	551	171
11620	5517	5517
	0.474785	

Bins	Random	Model
0	0	0
10%	551	1056
20%	1102	1998
30.0%	1653	2824
40.0%	2204	3518
50.0%	2755	4037
60.0%	3306	4412
70.0%	3857	4729
80.0%	4408	5057
90.0%	4959	5346
100.0%	5510	5517







Multi-class

• One versus all metrics

Any two class metrics

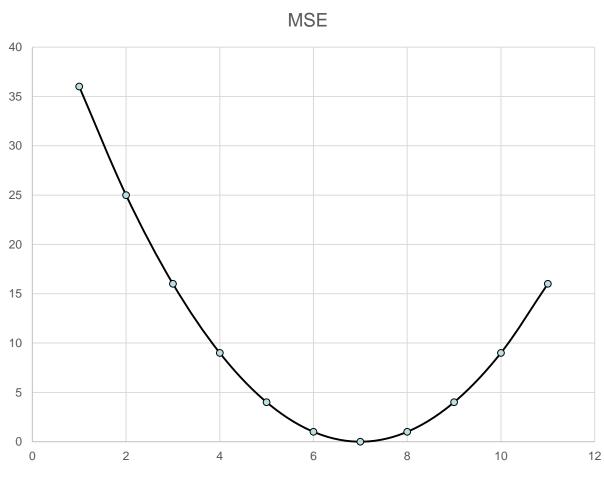
Your own metrics



FORECASTING



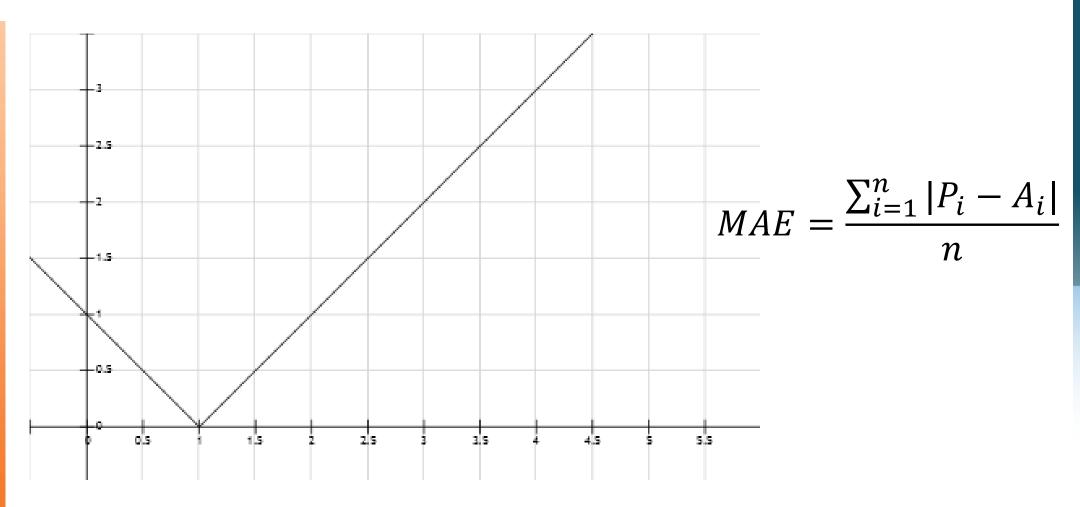
MSE(Mean square error)



$$MSE = \frac{\sum_{i=1}^{n} (P_i - A_i)^2}{n}$$

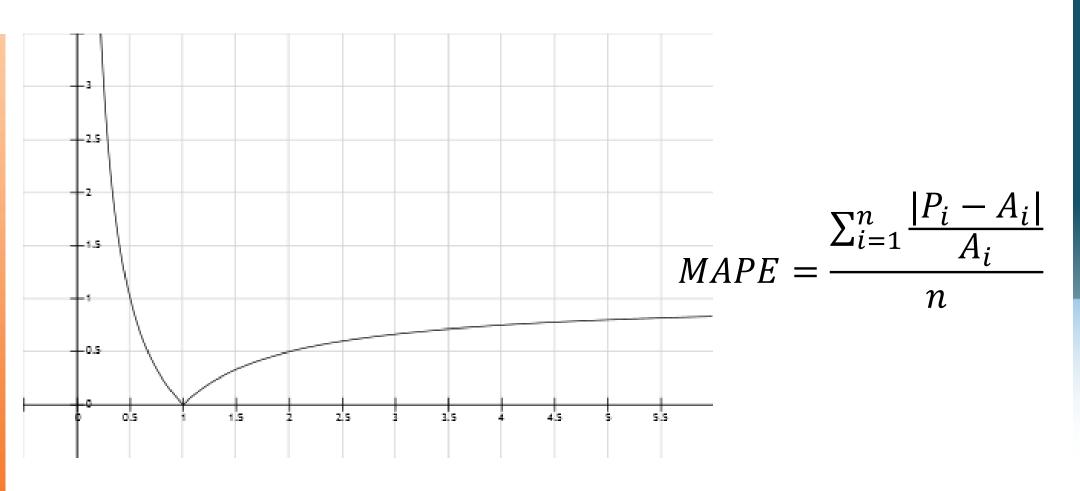


MAE (Mean absolute error)





MAPE(Mean absolute percentage error)





NMSE (Normalized Mean Square error)

$$NMSE = \frac{MSE \ of \ developed \ model}{MSE \ of \ naive \ model}$$



67



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