PRICE PREDICTOR

When is the best time to buy a product?

WHAT MACHINE LEARNING ALGORITHM SHOULD WE USE?

ARE THERE ANY GOOD METHODS FOR ENSURING OUR DATA REQUESTS DO NOT LOOK LIKE A DDOS ATTACKS?

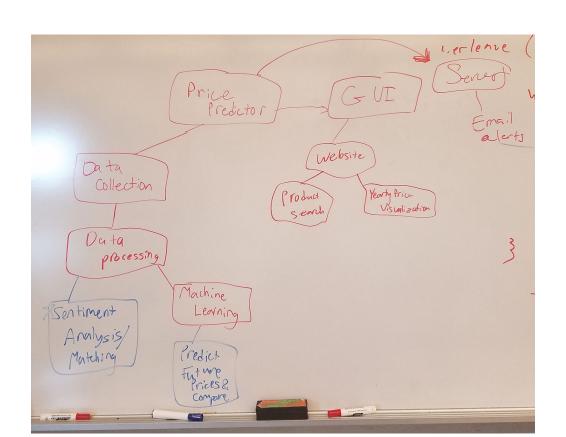
AGENDA

- Present our problems
- Brief overviews of potential solution algorithms
- Give sticky notes and ask peers to write down thoughts
- Have an mid presentation discussion with our peers about what machine learning algorithm appears best for our project
- Present our data collection technique
- Have discussion about DDOS attacks

WHAT OUR PROJECT IS:

- Scraping Amazon data for product price history
- Using sentiment analysis of reviews to help predict future prices
- Using machine learning for predicting the future prices
- Webapp where users search items and output when the best time to buy a product is
- Website that sends email when the best time to buy a product is

CODE ORGANIZATION AND CONCEPT MAP



WHY WE NEED YOU:

- We want help to decide which machine learning algorithms are best for our needs
- We want help ensuring our page requests do not look like DDOS attacks



MACHINE LEARNING ALGORITHM: LINEAR REGRESSION

 When an independent variable is used to predict the value of a dependent variable

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\circ Y = a + bx
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- Best at: Baseline predictions, economic predictions, modelling marketing responses
- Pros: fast to train, simple to understand, seldom overfits, easy to train on big data
- Cons: can suffer from outliers, hard to make it fit nonlinear functions

MACHINE LEARNING ALGORITHM: LEAST SQUARES REGRESSION

- Works like linear regression model, but minimizes the sum of the square of the differences between the observed and the predicted
- Advantages: works best if the trend is linear, maximum likelihood solution
- Disadvantages: unstable, highly influenced by outliers

MACHINE LEARNING ALGORITHM: RIDGE REGRESSION

- Ridge regression is like least squares but shrinks the estimated coefficients towards zero.
- Bias is introduced:
 - Bias increases as amount of shrinkage (lamda) increases
 - Variances decreases as amount of shrinkage (lamda) decreases
- Works well when subset of true coefficients are small

$$\sum_{i=1}^{n} \left(y_i - \beta_0 - \sum_{j=1}^{p} \beta_j x_{ij} \right)^2 + \lambda \sum_{j=1}^{p} \beta_j^2 = RSS + \lambda \sum_{j=1}^{p} \beta_j^2,$$

MACHINE LEARNING ALGORITHM: LASSO

- Least Absolute Shrinkage and Selection Operator
- Regression method that involves penalizing the absolute size of the regression coefficients
- Penalizing allows further estimates to go to zero
- Convenient when we want some automatic feature/variable selection or when dealing with highly correlated predictors

$$\min_{\beta} \frac{1}{n} (\mathbf{y} - X\beta)^T (\mathbf{y} - X\beta) + \lambda \sum_{j=1}^{d} |\beta_j|.$$

SINGULAR VALUE DECOMPOSITION

- Unsupervised Machine Learning
 - o no datasets are provided
- Can restructure data in a meaningful way
- A factorization of a matrix
 - \circ M = U Σ V, where U and V are unitary matrices and Σ is a diagonal matrix of singular values.
- Best for:
 - Recommender systems
- Difficult to understand why data has been restructured
 - Not very transparent

GATHERING DATA

- We are grabbing data from the Tracktor website which has Amazon price histories.
- We are reading from a specific part of the html in order to get the data.
- We want your help with methods we can use to make sure our html page requests do not look like a DDOS attack.

APPEARED DDOS ATTACKS?

- DDoS distributed denial-of-service
 - o attack occurs when multiple systems flood the bandwidth or resources of a targeted system, usually one or more web servers
 - Malicious
- Currently will not be a problem
 - We are not malicious
 - Educational purposes
 - May encounter this problem when scaling to more products