# Algorithms I Used



Apriori Algorythm Apriori $(T, \epsilon)$  $L_1 \leftarrow \{\text{large } 1 - \text{itemsets}\}$  $k \leftarrow 2$ while  $L_{k-1} \neq \emptyset$  $C_k \leftarrow \{a \cup \{b\} \mid a \in L_{k-1} \land b \notin a\} - \{c \mid \{s \mid s \subseteq c \land |s| = k-1\} \not\subseteq L_{k-1}\}$ for transactions  $t \in T$  $C_t \leftarrow \{c \mid c \in C_k \land c \subseteq t\}$ for candidates  $c \in C_t$  $count[c] \leftarrow count[c] + 1$  $L_k \leftarrow \{c \mid c \in C_k \land count[c] \ge \epsilon\}$  $k \leftarrow k + 1$ return  $\bigcup L_k$ Gini Index for  $I_G(f) = \sum_{i=1}^{m} f_i(1-f_i) = \sum_{i=1}^{m} (f_i - f_i^2) = \sum_{i=1}^{m} f_i - \sum_{i=1}^{m} f_i^2 = 1 - \sum_{i=1}^{m} f_i^2 = \sum_{i=1}^{m} f_i f_k$ decision tree grading

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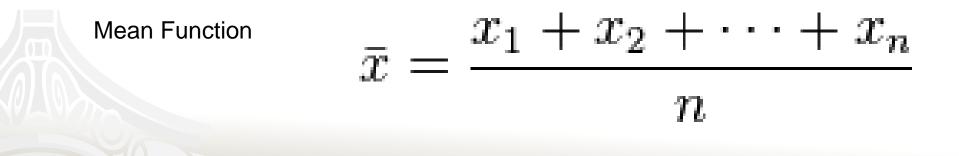


Euclidean Distance  $d(\mathbf{p}, \mathbf{q}) = d(\mathbf{q}, \mathbf{p}) = \sqrt{(q_1 - p_1)^2 + (q_2 - p_2)^2 + \dots + (q_n - p_n)^2}$ 

$$= \sqrt{\sum_{i=1}^{n} (q_i - p_i)^2}.$$

**Probability Function** 

$$P(X \le m) = P(X \ge m) = \int_{-\infty}^{m} f(x) dx = \frac{1}{2}.$$



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```
#Gave conditions where the value of SWAM would equal 1
SWAM$SMALL[grep1("S",SWAM$SWAM) | grep1("O",SWAM$SWAM)] <- 1
SWAM$WOMAN[grep1("W",SWAM$SWAM)] <- 1
SWAM$MINORITY[grep1("M",SWAM$SWAM)] <- 1
SWAM$MICRO[grep1("O",SWAM$SWAM)] <- 1
SWAM$SDV[grep1("SDV", SWAM$SWAM)] <- 1</pre>
```

#### Variable Generation

```
#Added attribute to allow for finding the amount of days in the fy that the order was published
BuyReq$DatesFromFYStart <- as.numeric(as.Date(as.character(BuyReq$PUBLISH_DATE),format="%m/%d/%Y")-as.Date(paste0(substr(
as.character(BuyReq$PUBLISH_DATE),nchar(as.character(BuyReq$PUBLISH_DATE))-3,nchar(as.character(BuyReq$PUBLISH_DATE))),"-07-01")))
%% 365
```

```
#made attribute flags to show what type of entity we are focusing on... School/Locality/Agency/ Other... other flag is used for
future purposes for now
BuyReg$IsSchool<-0;
BuyReg$IsSchool[grep1(SchoolIndicator, BuyReg$CLIENTNAME)] <-1;</pre>
BuyReg$IsLocality<-0;
BuyReq$IsLocality[grepl(LocalityIndicator,BuyReq$CLIENTNAME)] <-1;</pre>
BuyReg$IsOtherType<-0;
BuyReq$IsOtherType[grepl(OtherIndicator, BuyReq$CLIENTNAME)] <-1;</pre>
BuyReg$IsAgency <-0;
BuyReq$IsAgency[BuyReq$IsSchool==0 & BuyReq$IsLocality==0 & BuyReq$IsOtherType==0] <-1;
#When importing data from csv, a i.. sometimes appears, this code removes that... a better sql export would fix it too, utf-8-BOM
names(VenRes) <- gsub("i..", "", names(VenRes))</pre>
names(BuyReqLines) <- qsub("ï..", "", names(BuyReqLines))</pre>
names(VenResLines) <- gsub("i..", "", names(VenResLines))</pre>
names(BuyReq) <- gsub("i..", "", names(BuyReq))</pre>
names (BuyRegLines) [names (BuyRegLines) == "QUANTITY"] <- "QUANTITYrequested"
```

# Data Is Power



# Presented by Subhash Jaini

# FORUM2015

# Agenda



- Introduce myself
- Discuss the goal of analytics in eVA
- Questions



### Introduction



- Subhash Jaini
- Heavy background in restaurant management- passion for food
- Master Gardener
- New dad on May 7<sup>th</sup>, 2015
- Prior to eVA, Lead JavaScript Developer
- Evangelist for Data Science
- VCU's Executive Masters Decision Analytics
  - Logi Report Support/ Discover things

#### "Statistics Lie"



- eVA is VERY business oriented
- Logically driven, many meetings and debates for logic/data used/ business case
- Statistics/ Code/ Data is used to make the process replicable impersonally

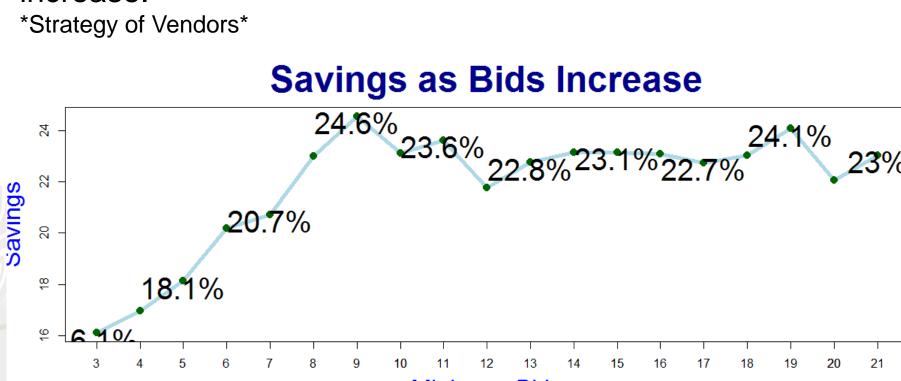
#### Analytics Goals: Here



Relevant

Trusted |Valuable

Through competition, we found proven cost savings. As more folks bid, the cost savings increase.



Here

**Minimum Bids** 

#### Analytics Goals: Relevant

Through heavier data engineering tactics, we were able to identify an item and run them through scenario analysis. Still experimental, but moving forward

Mandatory Contract vs Non Contract vs Non Sole Source vs No Self Registered vs Non

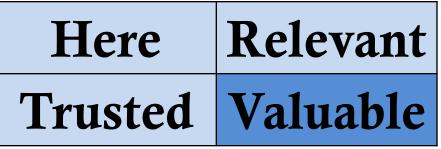


#### Analytics Goals: Trusted

Through PO data, we are able to identify an item and group it with other items based on a grouping. Groupings are based on orders, req, buyer AND day, BSO AND day. Still experimental, but neat stuff.

#### Analytics Goals: Valuable

- Personal Enjoyment
- •QQ Savings-Cost to Build
- •QQ Savings-Cost to Maintain
- QQ Savings-Savings to be had



#### Question Time



- Presentation: "How can professional buyers help in this data initiative?"
- Subhash's Turn
- Any Questions?



