Sales Forecasting using Data Mining

Heena Mansoori¹, Pooja Pawar² Computer Engineering^{1, 2,} Rajiv Gandhi Institute of Technology, Versova, Mumbai^{1, 2} Email: heena.mansoori@yahoo.in¹, p3crown@gmail.com²

Abstract- In today's dynamically evolving and changing retail industry, traditional business models and customer segments can no longer be taken for granted. Hence it is very important to leverage the numerous sources of data now readily available to the savvy business decision maker. This data can be used for business gain if the data is converted to information and then into knowledge. Timely identification of newly emerging trends in market is needed in business process. Sales patterns from inventory depicts market trends and can be used in forecasting which has great potential for decision making, strategic planning and market competition. In this paper we will analyze and implement data mining tools and techniques for sales forecasting then do an analysis of the sample or raw data to obtain a meaningful interpretation. Some of the data mining algorithms used in this paper are K-Means clustering algorithm, and then Most Frequent Pattern (MFP) to find frequencies of the items according to its sales and its property values.

Index Terms- Clustering, Most Frequent Pattern, Stock data, Decision Making, Data Mining

1. INTRODUCTION

After perusing different surveys on market's status, most executives realize that technology is paramount to driving growth and enhancing customer engagement. With unpredictable consumer behavior, spending, and demographic profiles changing enormously, organizations are finding it difficult to adopt a traditional approach. Therefore, retailers are perpetually embracing Data mining and analytics as key facilitators of success. Data Mining and analytics enable retailers to better understand customer profiles and extrapolate preferences by potentially predicting future customer purchases, assaying traffic patterns in stores, mining loyalty data, their geographical patterns, and monitoring consumer's behaviors. Retailers can dramatically enhance their connection to consumers and rely on Data Mining to identify new markets, new strategies, and new operational models to generate growth and profit. As it evolves in both sophistication and application, data mining and analytics will be leveraged by all departments of the retail organization.

2. EASE OF USE

2.1 Basket Analysis.

Combinations of products that frequently occur in transactions are found. For example, people who buy jacket in winter, also tend to buy socks (because a high proportion of them are planning to buy clothes for winter)

2.2 Sales Forecasting

Shows which product a customer buys during a particular season, and tries to predict when they will buy again. This type of analysis is used to determine when a product can remain obsolete.

2.3 Database Marketing

By assessing customer purchasing patterns and looking at the demographics of customers, one can build profiles, and create products resulting in more profit.

2.4 Merchandise Planning

Conducive for offline or online companies. For offline, a company looking to expand business by adding stores can evaluate the quantity and type of product for a particular season by looking at the graph generated for that item. For an online business, merchandise planning can help you determine stocking options and inventory warehousing

2.5 Market Segmentation

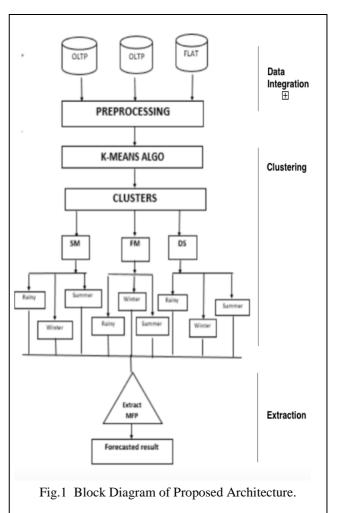
Database is segmented which is used to improve conversion rates so that the focus is on promotions for a tight, highly-interested market. It also helps to understand what factors are essential for growth by analyzing segments, allowing to customize products and promotions that satisfy the needs of that customers in a way a generic, broad promotion never will.

3. LITERARTURE REVIEW

The retail industry is more dynamic than ever hence retailers must evolve to succeed in the next decade.Data mining researchers constantly try to find most feasible and efficient methods for extraction of useful patterns from stock data.For stock data mining most researchers use the history of transactions as it likely that may persist in future. These can help to predict the customer behaviour and future trend. A model for inventory decision support system was implemented in which factors like quantity, cost, safety factor, lead time and backorder discounts were used as decision variables, the algorithm is applied to find the optimal solution for the case where the lead time demands follows a general distribution. Another proposed a pattern based stock data mining approach for intra-stock mining which focus on finding frequently appearing patterns for the stock time series data and inter stock mining which discover the strong relationship among the several stocks.

Recently, it has been recognized that the partitioned clustering technique is well suited for clustering a large dataset the best well known clustering algorithm is the K-means algorithm and its variants. This algorithm is simple, reliable, and straightforward and is based firmly on foundation of analysis of variances. In addition to the K-means algorithm, other algorithms, like Particle Swarm Optimization (PSO) is another computational intelligence method which was applied for image clustering and other low dimensional datasets. For this project, we have used clustering algorithm to form clusters of data and MFP to find association patterns amongst them. The classification of similar objects into several groups, or partitioning of data into subsets called clusters. Data in each subset share some common proximity value accordingly to some defined distance measure.

In this work we have proposed an algorithm for mining patterns of huge stock data to show factors affecting the sale of products. By using k-means clustering algorithm we form clusters of data based on its sold quantities for each category of items i.e. Dead-Stock (DS), Slow-Moving (SM) and Fast- Moving (FM).Most Frequent Pattern (MFP) algorithm to find frequencies of property values of the corresponding items. MFP provides frequent patterns of items in each category of products. This work is similar to Apriori algorithm used to find strong association patterns which are visible. From the experimental results extracted from sample data, it is illustrated that the proposed algorithm of k-mean and association of MFP can generate more useful and beneficent patterns from large stock data.



4. PROPOSED SYSTEM

In this paper we have proposed an algorithm for mining patterns of huge stock data to show the factors affecting the sales of product. In the first phase, we divide the stock data into nine different clusters categorically using k-means algorithm. On the basis of its sold quantity, each category of items is formed which are named as Dead-Stock (DS), Slow-Moving (SM) and Fast- Moving (FM) and the season in which they were sold i.e. Winter, Summer and Rainy .Moving with second phase we have proposed Most Frequent Pattern (MFP) algorithm which is used to extract frequent patterns of item attributes from each category of products. Cluster analysis techniques are widely used in market research when working with data from various surveys and test panels. Market researchers use cluster analysis to partition the population of consumers to form market segmentation, Intra and inter stock patterns and to understand better association patterns between them.

4.1 K-Means

K-means is a typical clustering algorithm and has been used for classification of data for decades. In this approach the data objects are classified on the basis of their attributes into k number of clusters. Euclidean distance is used to calculate proximity between clusters.

$$dist(i, j) = \sqrt{\sum_{k=1}^{l} (x_{ik} - x_{jk})^2}$$

This algorithm aims at minimizing an objective function, in this case a squared error function. The objective function is

$$J = \sum_{j=1}^{k} \sum_{i=1}^{k} \left\| \boldsymbol{x}_{i}^{(j)} - \boldsymbol{c}_{j} \right\|^{2}$$

Where
$$\left\| \boldsymbol{x}_{i}^{(j)} - \boldsymbol{c}_{j} \right\|^{2}$$

Is a chosen distance is measure between a data point X i(j) and the cluster center c j, is an indicator of the distance of the n data points from their respective cluster centers. The steps of the K-means algorithm are as described below

- Place K points into the space represented by the objects that are being clustered; these points represent initial group centroids.
- Assign each object to the group that has the closest centroid
- When all objects have been assigned, recalculate the positions of the K centroids.

Repeat Steps 2 and 3 until the centroids no longer move. This produces a separation of the objects into groups from which the metric to be minimized can be calculated

Cluster3. Dead	Stock Winter					
Cluster Id	Product Name	Color	Gender	Season	Qty Sold	month
1	T-shirt	Blue	Female	Winter	46	Novemb
2	Coat	Blue	Male	Winter	44	Novemb
3	Coat	White	Male	Winter	44	Februa
4	Sweater	White	Male	Winter	42	Decemt
5	Sweater	White	Female	Winter	34	Februar
8	Shoes	Red	Male	Winter	32	Novemb
7	T-shirt	White	Male	Winter	28	Novemb
8	Jeans	Blue	Male	Winter	28	Februar
				Winter		
8	T-shirt.	Green	Female		26	Novemb
10	T-shirt	Red	Male	Winter	24	Novemb
			12			
Cluster4. Slow	Moving Rainy					
Cluster Id	Product Name	Color	Gender	Season	Qty Sold	month
1	Jeans	Black	Male	Rainy	86	August
2	Shirt	Blue	Male	Rainy	78	August
3	T-shirt	Red	Male	Rainy	60	July
4	Jeans	Blue	Female	Rainy	60	Septemb
5	T-shirt	Green	Female	Rainy	56	August
6	Coat	Red	Female	Rainy	48	August
7	Coat	Red	Female	Rainy	48	July
	Moving Summer					
Cluster Id	Product Name	Color	Gender	Season	Qty Sold	
<u>Cluster Id</u> 1	Product Name Shoes	White	Male	Summer	78	Ap
<u>Cluster Id</u> 1 2	Product Name Shoes T-shirt	White Yellow	Male Female	Summer Summer	78 74	Ap Ju
<u>Cluster Id</u> 1 2 3	Product Name Shoes T-shirt T-shirt	White Yellow White	Male Female Male	Summer Summer Summer	78 74 68	Ap Ju Ma
Cluster Id 1 2 3 4	Product Name Shoes T-shirt T-shirt T-shirt	White Yellow White White	Male Female Male Female	Summer Summer Summer Summer	78 74 68 60	Ap Jur Ma Ma
Cluster Id 1 2 3 4 5	Product Name Shoes T-shirt T-shirt T-shirt T-shirt	White Yellow White White Green	Male Female Male Female Male	Summer Summer Summer Summer Summer	78 74 68 60 60	Ap Jur Ma Ma Ap
<u>Cluster Id</u> 1 2 3 4 5 5	Product Name Shoes T-shirt T-shirt T-shirt T-shirt Jeans	White Yellow White White Green Blue	Male Female Male Female Male Female	Summer Summer Summer Summer Summer Summer	78 74 68 60 60 60	Ap Ju Mi Mi Ap Ap
<u>Cluster Id</u> 1 2 3 4 5 5 5 7	Product Name Shoes T-shirt T-shirt T-shirt Jeans Jeans	White Yellow White White Green Blue Blue	Male Female Male Female Male	Summer Summer Summer Summer Summer	78 74 68 60 60 60 60 54	Ap Jur Ma Ma Ap Ap Ap
<u>Cluster Id</u> 1 2 3 4 5 5	Product Name Shoes T-shirt T-shirt T-shirt T-shirt Jeans	White Yellow White White Green Blue	Male Female Male Female Male Female	Summer Summer Summer Summer Summer Summer	78 74 68 60 60 60 60 54	Ap Jun Ma Ma Ap Ap Ap
<u>Cluster Id</u> 1 2 3 4 5 5 5 7	Product Name Shoes T-shirt T-shirt T-shirt Jeans Jeans Shoes	White Yellow White White Green Blue Blue	Male Female Male Female Male Female Male	Summer Summer Summer Summer Summer Summer	78 74 68 60 60 60 60 54	Ap Jur Ma Ma Ap Ap Ap
<u>Cluster Id</u> 1 2 3 4 5 5 5 7 8	Product Name Shoes T-shirt T-shirt T-shirt Jeans Jeans Shoes	White Yellow White White Green Blue Blue	Male Female Male Female Male Female Male	Summer Summer Summer Summer Summer Summer	78 74 68 60 60 60 60 54	Ар Ји М: М Ар Ар Ар
<u>Cluster Id</u> 1 2 3 4 5 5 7 8 Cluster6. Slow	Product Name Shoes T-shirt T-shirt T-shirt Jeans Jeans Shoes Moving Winter	White Yellow White White Green Blue Blue White	Male Female Male Female Female Male Female	Summer Summer Summer Summer Summer Summer Summer	78 74 88 60 60 60 54 48	Ap Ju Mi Ap Ap Ap Ap
<u>Cluster Id</u> 1 2 3 4 5 5 7 8 Cluster Slow <u>Cluster Id</u>	Product Name Shoes T-shirt T-shirt T-shirt Jeans Jeans Shoes Moving Winter Product Name	White Yellow White White Blue Blue White <u>Color</u>	Male Female Male Female Male Female Female	Summer Summer Summer Summer Summer Summer Summer	78 74 88 60 80 60 54 48 Qty Sold	Aş Ju Mı Aş Aş Aş Aş Decemt
Cluster Id 1 2 3 4 5 5 7 8 Cluster Slow Cluster Id 1	Product Name Shoes T-shirt T-shirt T-shirt Jeans Jeans Shoes Moving Winter Product Name Shoes	White Yellow White Green Blue Blue White Color Black	Male Female Male Male Female Female Giender Male	Summer Summer Summer Summer Summer Summer Summer Summer	78 74 88 60 80 50 54 48 9ty Sold 88	Ap Ju Mi Ap Ap Ap Ap Ap Ap Novemb
Cluster Id 1 2 3 4 5 6 7 8 Cluster6. Slow Cluster Id 1 2	Product Name Shoes T-shirt T-shirt T-shirt T-shirt Jeans Jeans Shoes Moving Winter Product Name Shoes Shirt	White Yellow White Green Blue Blue White Color Black Red	Male Female Male Female Male Female Female Gender Male Male	Summer Summer Summer Summer Summer Summer Summer Winter Winter	78 74 88 60 60 60 54 48 <u>Qty Sold</u> 88 80	Ap Ju Mi Ap Ap Ap Ap Ap Novemb
Cluster Id	Product Name Shoes T-shirt T-shirt T-shirt Jeans Jeans Shoes Moving Winter Product Name Shoes Shirt Sweater	White Yellow White Green Blue Blue White Color Black Red Blue	Male Female Male Female Male Female Male Male Female	Summer Summer Summer Summer Summer Summer Summer Winter Winter Winter	78 74 88 60 60 54 48 9ty Sold 88 80 74	Ap Ju Mi Ap Ap Ap Ap Decemb Novemb Novemb
Cluster Id	Product Name Shoes T-shirt T-shirt T-shirt Jeans Jeans Shoes Moving Winter Product Name Shoes Shirt Sweater T-shirt	White Yellow White White Green Blue Blue White Color Black Red Blue White	Male Female Male Female Male Female Gender Male Female Female	Summer Summer Summer Summer Summer Summer Summer Winter Winter Winter Winter	78 74 88 60 60 60 54 48 88 88 80 74 74	Ap Ju Mi Ap Ap Ap Ap Decemb Novemb Novemb Januar Novemb
Cluster Id 1 2 3 4 5 5 5 7 8 Cluster6. Slow Cluster Id 1 2 3 4 5 5 5 5 5 5 7 8 Cluster6. Slow 5 5 5 5 5 5 5 5 5 5 5 5 5	Product Name Shoes T-shirt T-shirt T-shirt Jeans Jeans Jeans Shoes Moving Winter Product Name Shoes Shirt Sweater T-shirt Shirt	White Yellow White Green Blue Blue White Black Red Blue White Black	Male Female Male Female Male Female Male Female Female Female Male	Summer Summer Summer Summer Summer Summer Summer Winter Winter Winter Winter Winter	78 74 88 60 60 60 54 48 88 80 74 74 72	Ap Ju Mi Ap Ap Ap Ap Decemb Novemb Novemb Januar Novemb Decemb
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4.2. MOST FREQUENT PATTERN

Association rule mining is a well-defined and renowned technique used for extracting correlations, frequent patterns, associations among sets of item in Relational databases, transactional databases or other information repositories. Association rules are widely used in areas such as risk management, telecommunications, market basket analysis, inventory control, and in medical field. Apriori algorithm is the core of various algorithms used in data mining. It analyzes a dataset for determining which combinations of products occur together frequently .MFP efficiently generates frequent patterns and strong association between data used. For this purpose a matrix containing counted values of corresponding properties of each product has been used.

	MFPColorMatrix_Id	Product_Name	Color	SaleCount
1	1	Coat	Black	178
2	2	Jeans	Black	184
3	3	Shirt	Black	114
4	4	Shoes	Black	160
5	5	Sweater	Black	20
6	6	T-shirt	Black	76
7	7	Coat	Blue	110
8	8	Jeans	Blue	338
9	9	Shirt	Blue	242
10	10	Shoes	Blue	44
11	11	Sweater	Blue	246

	MFPGenderMatrix_ld	Product_Name	Gender	GenderCount
1	1	Coat	Female	218
2	2	Jeans	Female	238
3	3	Shirt	Female	145
4	4	Shoes	Female	170
5	5	Sweater	Female	396
6	6	T-shirt	Female	484
7	7	Coat	Male	328
8	8	Jeans	Male	418
9	9	Shirt	Male	400
10	10	Shoes	Male	352
11	11	Sweater	Male	76

In the tables shown above we have generated MFP results by using count of attributes along with its properties from Clustered Data Results and accordingly formed 3 matrixes on the basis of Color, Gender and Season.

	MFPSeasonMatrix_Id	Product_Name	Season	SeasonCount
£	1	Coat	Rainy	140
	2	Jeans	Rainy	188
	3	Shirt	Rainy	162
ŧ.	4	Shoes	Rainy	78
5	5	Sweater	Rainy	44
5	6	T-shirt	Rainy	242
1	7	Coat	Summer	108
	8	Jeans	Summer	312
•	9	Shirt	Summer	90
0	10	Shoes	Summer	198
1	11	Sweater	Summer	10

5. RESULTS

This is the final forecasted results showing following values

- MFP id: This tab gives a unique item id for each product item present in the Dataset. For experimental purpose initially 15 product items used for three seasons i.e. Rainy, Winter and Summer
- Product Name: This tab lists down all the available product items on the basis of which these products are classified Color wise, Gender wise and Season wise
- Color: This tab gives the values of all the available colors of product items.
- Color_count: This tab gives the total number of color counts for that product sold.
- Gender: This tab helps the admin to identify the product purchased by whom either by a male or a female customer. This is an important parameter to be known for decision making and launching product in market.
- Gender_count: This tab gives the number of customers according to gender to identify the product sold.
- Season: This tab corresponds to the season in which that product was sold. This too helps the admin to identify the time to launch that product and forecast easily.
- Season_Count: This tab gives the total number of products sold in that season.

Hence by analyzing these results, retailers can understand what is currently trending in market, which product is preferred more by customers and during which season. By careful observation, retailers can predict what will be the customer behavior in future and can devise sales plan accordingly resulting in profits.

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MFP_ld	Product_Name	Color	Color_Count	Gender	Gender_Count	Season	Season_Count
1	Coat	Black	876	Female	1117	Winter	841
2	Jeans	Black	562	Male	995	Summer	759
3	Shirt	Black	570	Male	1186	Winter	702
4	Shirts	White	8	Female	145	Summer	130
5	Shoes	Black	783	Male	1189	Summer	776
6	Sweater	Blue	652	Female	1111	Winter	1487
7	T-Shirt	Blue	626	Male	1555	Summer	831
8	T-shirts	Black	41	Male	131	Rainy	172

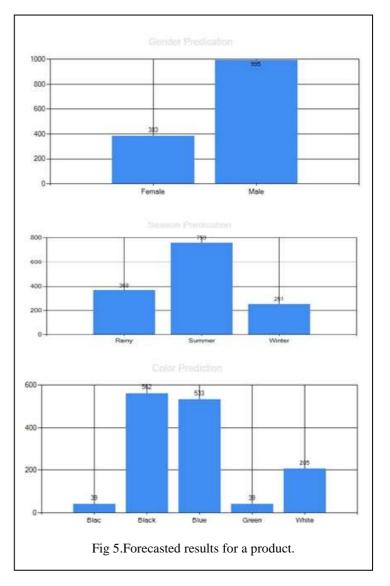
Available online at www.ijrat.org

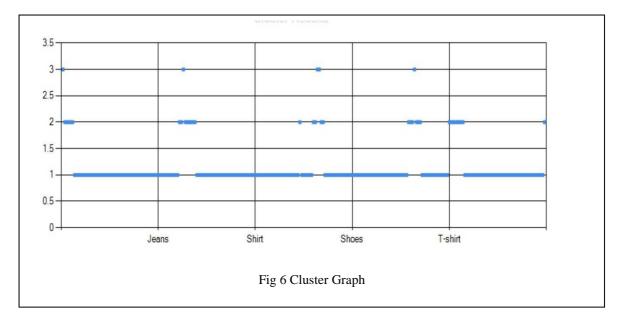
The graph shown in Fig 5 shows Color wise, Gender wise and Season-wise prediction for product coat.

6. CONCLUSION

In this paper, the problem of pattern recognition from stock data is addressed. Hybrid clustering association mining approach is proposed to classify stock data and find a more compact associated pattern of sales. From the results it is clearly evident that the proposed approach is very efficient and highly reliable for mining patterns of huge stock data and predicting the factors affecting the sales of product.

We articulated most frequent pattern of products by using their properties from the inventory system. We assimilated the trends of selling products by meticulously analyzing the forecasted results formed by using Product matrix. The limitation of study is, that it requires a data format with specific attributes. In future we will extend our work to implement at the preprocessing stage where the system loads a single dataset file and applies classification algorithms; here we can add multiple dataset files for simultaneous operations like cluster analysis, MFP and forecasting result with final result graphs. We can add this functionality which will make this application capable to extract more precise information in large amount.





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