

Recommendation platform





Problem Statement

- Creating a product recommendation system
- With the help of Machine Learning increase sales
- Reduce **time** spent by Sales Representatives





Solution

- Create a **Product to Store** based recommendation system that is Centrally controlled
- Recommendations made based on previous sales history of the product
- Integrate Online sales trends to provide better real time product recommendation



Recomendation engine

- Product recommendations are made separately for each and every store depending on the sales history
- Features include quantity of products sold, time of the year (monthly or weekly)
- Location based features such as temperature and climate



Recomendation engine contd.

- Products **sold together** are recommended to the store by the distributor
- Integration with e-commerce websites to provide a real time trend analysis
- This will help analyse market position for each product separately





Recomendation engine contd.

- Online reviews work as a **bonus factor** to promote products
- Incase of poor online performance of a product, it provides a range of factors to **grade** the **competition**
- Integrating this complete platform will provide an edge over current traditional methods





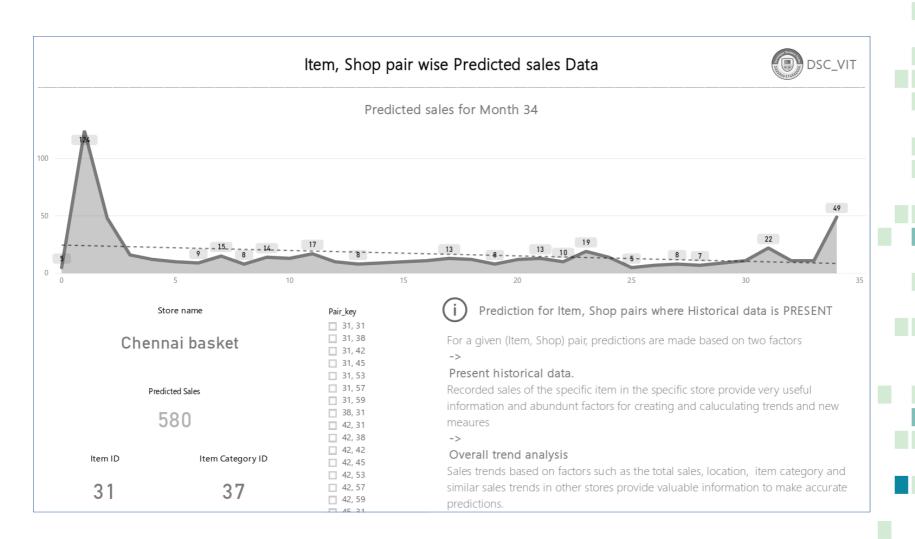
Expected output

- Recommendation system with store based product recommendation
- Distributors get sales information on each and every product sold to the store
- Overall sales as a Key Performance Indicator (KPI)





Report





Notebooks

```
In [27]: # Enter product ID to get a list of 20 recommended items
         # User entered value
         product id = 'B00L5JHZJ0'
         product data = [data model norm.loc[product id][['Count', 'Rating', 'Unreliability']].values]
         recommended products = engine.kneighbors(X=product data, n neighbors=20, return distance=False)
         # List of product IDs form the indexes
         products list = []
         for each in recommended products:
             products list.append(data model norm.iloc[each].index)
         print("Recommended products: ")
         print(products list)
         # Showing recommended products
         ax = data model norm.plot(kind='scatter', x='Rating', y='Count', color='grey', alpha=0.20)
         data model norm.iloc[recommended products[0]].plot(kind='scatter', x='Rating', y='Count',\
                                                            color='orange', alpha=0.5, ax=ax)
         ax2 = data model norm.plot(kind='scatter', x='Rating', y='Unreliability', color='grey', alpha=0.20)
         data model norm. iloc[recommended products[0]].plot(kind='scatter', x='Rating', y='Unreliability',\
                                                             color='orange', alpha=0.5, ax=ax2)
         Recommended products:
         [Index([u'B00L5JHZJ0', u'B00021C1LI', u'B00FAE0CP0', u'B00178TVXG',
                u'B000ZLVUYO', u'B00804YM4Y', u'B000L9YYTS', u'B001MWV40U',
                u'B000F63TW0', u'B0000Q2DL4', u'B001330XFA', u'B001HTYJL0',
                u'B000TAE0RK', u'B0023A7JF4', u'B001TJXI5U', u'B008RVYJS8',
                u'B0008IV7BU', u'B001C6H6F8', u'B000G33KE0', u'B000GHWSG6'],
               dtype='object', name=u'ProductId')]
Out[27]: <matplotlib.axes._subplots.AxesSubplot at 0x7f9fa8552110>
            10
                                  Rating
```



THANKS!

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Any questions?

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