

```
In [1]: import pandas as pd

In [7]: df=pd.read_csv('Inventory_Stock_Data.csv')

In [8]: df.shape
Out[8]: (118, 11)

In [35]: #for numerical columns for inventory_stock_delivery
df[['product id','avg lead time','avg order qty','current stock','max lead time','max order qty','mod','reorder point','safety stock']]

Out[10]:
   product id  avg lead time  avg order qty  current stock  max lead time  max order qty  mod  reorder point  safety stock
0           0           19           10           1           69           30           1           1           35           25
1           1           24           10           3           175           30           5           0           175           144
2           2           35           10           1           65           30           1           1           35           25
3           3           37           10           3           328           30           5           1           175           146
4           4           44           10           3           175           30           5           0           175           145
...         ...         ...         ...         ...         ...         ...         ...         ...         ...         ...
113          113          1959          10           1           69           30           1           1           35           25
114          114          1360          10           1           35           30           1           0           35           25
115          115          1361          10           1           61           30           1           1           35           25
116          116          1362          10           1           35           30           1           0           35           25
117          117          1363          10           1           63           30           1           1           35           25

118 rows x 9 columns

In [10]: #for categorical columns for inventory_stock_delivery
df[['order-now','product name']]

Out[10]:
   order-now  product name
0         green  Nike Men's Fingtrap Max Training Shoe
1         orange  Elevation Training Mask 2.0
2         green  adidas Brazuca 2014 Official Match Ball
3         green  adidas Kids' F5 Messi FG Soccer Cleat
4         orange  adidas Mens F10 Messi TRX FG Soccer Cleat
...         ...         ...
113         green  Adult dog supplies
114         orange  Smart watch
115         green  Toys
116         orange  Fighting video games
117         green  Summer dresses

118 rows x 2 columns

In [38]: df1=pd.read_csv('Sales_Shipment_Data.csv')

In [39]: df1.shape
Out[39]: (180519, 46)

In [24]: #for numerical columns for Sales_Shipment_Data
df1[['Department Id','Order Id','Product Id','shipping date (DateOrders)','Benefit per order','Days for shipment (scheduled)','Days for shipping (real)','Latitude',

Out[24]:
   Department Id  Order Id  Product Id  shipping date (DateOrders)  Benefit per order  Days for shipment (scheduled)  Days for shipping (real)  Latitude  Longitude  Order Customer Id  ...  Order Item Discount  Order Item Product Price  Order Item Profit Ratio  Order Item Quantity  Order Item Total  Order Profit Per Order  Order Item Total
0              5    53963      502      3/2/2017 17:17  9.610000          2          4    33.425201  -104.522080    5673  ...  15.000000          0.15    134920  50.000000          0.11          2      85.000000
1              1     53908      502  3/30/2015 0:47  9.490000          2          4    34.034836  -118.260025    9723  ...  16.000000          0.16    9722  50.000000          0.11          2      84.000000
2              5    52009      502  2/3/2017 4:43  30.709999          2          5    41.623779  -87.648697    9511  ...  5.500000          0.06   129954  50.000000          0.33          2      84.500000
3              5     51179      502    1/23/2015 4:42  27.549999          2          5    41.904102  -87.937843    3326  ...  5.000000          0.05   2929  50.000000          0.29          2      95.000000
4              5    58019      502    4/2/2017 17:36  47.500000          2          5    36.225159  -115.195183    907   ...  5.000000          0.05   140056  50.000000          0.50          2      85.000000
...         ...         ...         ...         ...         ...         ...         ...         ...         ...         ...         ...         ...         ...         ...         ...         ...         ...
180514         4     48890      365  1/4/2017 6:20 -302.350006          4          6    33.320156  -117.156357    9615  ...  47.990002          0.16   124746  59.990002          -1.20          5    251.960007
180515         4     28410      365    3/11/2016 7:16  117.010002          4          6    47.678699  -122.321762    1971  ...  50.990002          0.17    73577  59.990002          0.47          5    248.960007
180516         3     24350      191    12/28/2015 10:32  -46.869999          4          6    33.823872  -118.007393    7663  ...  124.989998          0.25   60910  99.989998          -0.13          5    374.959992
180517         4     45725      365    11/4/2016 11:09  125.260002          4          6    32.719875  -117.139725    415  ...  38.990002          0.13   114288  59.990002          0.48          5    260.959992
180518         4     24050      365    12/24/2015 1:26  117.010002          4          6    38.495419  -121.479187    7341  ...  50.990002          0.17   60174  59.990002          0.47          5    248.960007

180519 rows x 21 columns

In [25]: #for categorical columns for Sales_Shipment_Data
df1[['Category Name','Class','Customer City','Customer Country','Customer Fname','Customer Lname','Customer Segment','Customer State','Customer Street','Delivery Status',

Out[25]:
   Category Name  Class  Customer City  Customer Country  Customer Fname  Customer Lname  Customer Segment  Customer State  Customer Street  Delivery Status  Department Name  Market  Order City  Order Country  Order Region  Order State  Order State
0  Women's Apparel  Moderate Value-Moderate Number  Roswell  EE.UU.  Jean  Dunn  Corporate  NM  8837 Blue Hills Harbour  Late  Golf LATAM  LATAM  Tlalpan  México  Central America  Oaxtla Federal  Oaxtla
1  Women's Apparel  Moderate Value-Moderate Number  Los Angeles  EE.UU.  Rose  White  Corporate  CA  1396 Silver Berry Impasse  Late  Golf LATAM  LATAM  Soyapango  El Salvador  Central America  San Salvador  Pinar
2  Women's Apparel  Moderate Value-Moderate Number  Harvey  EE.UU.  Joshua  Brady  Corporate  IL  8414 Wishing Horse Acres  Late  Golf LATAM  LATAM  Managua  Nicaragua  Central America  Managua  Pinar
3  Women's Apparel  Moderate Value-Moderate Number  Elmhurst  EE.UU.  Mary  Smith  Corporate  IL  1269 Tawny Key  Late  Golf LATAM  LATAM  Poza Rica de Hidalgo  México  Central America  Veracruz  Pinar
4  Women's Apparel  Moderate Value-Moderate Number  North Las Vegas  EE.UU.  Samantha  Kline  Corporate  NV  6396 Blue Gate Parkway  Late  Golf LATAM  LATAM  San Pedro Sula  Honduras  Central America  Cortés  Pinar
...         ...         ...         ...         ...         ...         ...         ...         ...         ...         ...         ...         ...         ...         ...         ...         ...
180514  Cleats  High Value-Small Number  Escondido  EE.UU.  Charles  Smith  Corporate  CA  4088 Stony Vista  Late  Apparel  Pacific Asia  Haifa  Israel  West Asia  Haifa  Comp
180515  Cleats  High Value-Small Number  Seattle  EE.UU.  Richard  Stokes  Corporate  WA  2865 Blue Rice Way  Late  Apparel  Pacific Asia  Yuyao  China  Eastern Asia  Zhejiang  Comp
180516  Cardio Equipment  High Value-Small Number  Anaheim  EE.UU.  Mary  Smith  Corporate  CA  2792 Story Range  Late  Footwear  Pacific Asia  Rangún  Myanmar (Birman)  Southeast Asia  Rangún  On F
180517  Cleats  High Value-Small Number  San Diego  EE.UU.  Mary  Allen  Corporate  CA  3983 Golden By-pass  Late  Apparel  Pacific Asia  Kahramanmara?  Turgula  West Asia  Kahramanmara?  On F
180518  Cleats  High Value-Small Number  Sacramento  EE.UU.  Mary  Jackson  Corporate  CA  7760 Golden Limits  Late  Apparel  Pacific Asia  Jiangmen  China  Eastern Asia  Guangdong  On F

180519 rows x 20 columns

In [ ]: import pandas as pd

In [ ]: import numpy as np

In [ ]: df=read_csv('Inventory_Stock_Data.csv')

In [ ]: df.shape
Out[ ]: (118, 11)

In [1]: import pandas as pd

In [2]: import numpy as np

In [3]: df=pd.read_csv('Inventory_Stock_Data.csv')

In [7]: sum_cols=[ 'product id','avg lead time','avg order qty','current stock','max lead time','max order qty','mod','reorder point','safety stock']
df(sum_cols).sum()

Out[7]:
product id      81370
avg lead time    1180
avg order qty     246
current stock   18965
max lead time    3540
max order qty     374
mod              64
reorder point   13890
safety stock    18659
dtype: int64

In [ ]:

In [6]: c

In [7]: import numpy as np

In [8]: df1=pd.read_csv('Sales_Shipment_Data.csv')

In [9]: df1.shape
Out[9]: (180519, 46)

In [10]: sum_cols=[ 'Department Id','Order Id','Product Id','shipping date (DateOrders)','Benefit per order','Days for shipment (scheduled)','Days for shipping (real)',
df1(sum_cols).sum()

Out[10]:
Department Id      982648
Order Id          6538740246
Product Id        129011170
shipping date (DateOrders)  3/2/2017 17:17:3/3/2015 0:47:2/3/2017 4:43:1/23/2...
Benefit per order      3966902.97405
Days for shipment (scheduled)  529254
Days for shipping (real)      631383
Latitude          5365016.495643
Longitude        -15328892.678974
Order Customer Id    1297021135
Order Item Cardprod Id  125011170
Order Item Discount      3738376.462593
Order Item Profit Ratio  18353.64669
Order Item Product Price  16293644940
Order Item Profit Per Order  25495158.688679
Order Item Quantity      21779.809992
Order Item Total      384079
Order Profit Per Order  33054482.380217
Product Price      3966902.97405
Sales per customer      25495158.688679
dtype: object

In [14]: df1['Order status'].value_counts()['Complete']
Out[14]: 59491

In [3]: import pandas as pd

import numpy as np

In [4]: df1=pd.read_csv('Sales_Shipment_Data.csv')

In [5]: df1.shape
Out[5]: (180519, 46)

In [7]: df1['Delivery Status'].value_counts()['Late']
Out[7]: 98977

In [8]: df1['Delivery Status'].value_counts()['Advance']
Out[8]: 41592

In [9]: df1['Delivery Status'].value_counts()['Canceled']
Out[9]: 7754

In [10]: df1['Delivery Status'].value_counts()['On time']
Out[10]: 32196

In [13]: import pandas as pd

In [14]: import numpy as np

In [ ]:

In [17]:

In [18]:

In [ ]:

In [ ]:

In [ ]:

In [2]: import pandas as pd

In [3]: import numpy as np

In [4]: df1=pd.read_csv('Sales_Shipment_Data.csv')

In [5]: df1[date_time]=[['order date (DateOrders)','shipping date (DateOrders)']]

In [9]: date_time

Out[9]:
   order date (DateOrders)  shipping date (DateOrders)
0      2/26/2017 17:17      3/2/2017 17:17
1      2/27/2015 0:47      3/30/2015 0:47
2      1/29/2017 4:43      2/3/2017 4:43
3      1/18/2015 4:42      1/23/2015 4:42
4      3/28/2017 17:36      4/2/2017 17:36
...         ...         ...
180514  12/29/2016 6:20      1/4/2017 6:20
180515  3/5/2016 7:16      3/11/2016 7:16
180516  12/22/2015 10:32      11/28/2015 10:32
180517  10/29/2016 11:09      1/4/2016 11:09
180518  12/18/2015 1:26      12/24/2015 1:26

180519 rows x 2 columns

In [38]: df1['shipping date (DateOrders)']=pd.to_datetime(df1['shipping date (DateOrders)'])

In [39]: df1['shipping date (DateOrders)']

Out[39]:
0      2017-03-02 17:17:00
1      2015-03-03 00:47:00
2      2017-02-03 04:43:00
3      2015-01-23 04:42:00
4      2017-04-02 17:36:00
...
180514  2017-01-04 06:20:00
180515  2016-03-11 07:16:00
180516  2015-12-28 10:32:00
180517  2016-11-04 11:09:00
180518  2015-12-24 01:26:00
Name: shipping date (DateOrders), Length: 180519, dtype: datetime64[ns]

In [2]: import pandas as pd

import numpy as np

In [3]: df1=pd.read_csv('Sales_Shipment_Data.csv')

In [4]: df1

Out[4]:
   Product Category Name  Class  Customer City  Customer Country  Customer Fname  Customer Lname  Customer Segment  Customer State  Order Item Discount Rate  Order Item Id  Order Item Product Price  Order Item Profit Ratio  Order Item Quantity  Order Item Total  Order Profit Per Order  Product Price  Sales
0           Women's Apparel  Moderate Value-Moderate Number  Roswell  EE.UU.  Jean  Dunn  Corporate  NM  ...  0.15    134920  50.000000          0.11          2      85.000000          9.610000      50.0
1           Women's Apparel  Moderate Value-Moderate Number  Los Angeles  EE.UU.  Rose  White  Corporate  CA  ...  0.16    9722  50.000000          0.11          2      84.000000          9.490000      50.0
2           Women's Apparel  Moderate Value-Moderate Number  Harvey  EE.UU.  Joshua  Brady  Corporate  IL  ...  0.06   129954  50.000000          0.33          2      84.500000          30.709999      50.0
3           Women's Apparel  Moderate Value-Moderate Number  Elmhurst  EE.UU.  Mary  Smith  Corporate  IL  ...  0.05   2929  50.000000          0.29          2      95.000000          27.549999      50.0
4           Women's Apparel  Moderate Value-Moderate Number  North Las Vegas  EE.UU.  Samantha  Kline  Corporate  NV  ...  0.05   140056  50.000000          0.50          2      85.000000          47.500000      50.0
...         ...         ...         ...         ...         ...         ...         ...         ...         ...         ...         ...         ...         ...         ...         ...         ...
180514  Cleats  High Value-Small Number  Escondido  EE.UU.  Charles  Smith  Corporate  CA  ...  0.16   124746  59.990002          -1.20          5    251.960007          -302.350006      59.9
180515  Cleats  High Value-Small Number  Seattle  EE.UU.  Richard  Stokes  Corporate  WA  ...  0.17    73577  59.990002          0.47          5    248.960007          117.010002      59.9
180516  Cardio Equipment  High Value-Small Number  Anaheim  EE.UU.  Mary  Smith  Corporate  CA  ...  0.25   60910  99.989998          -0.13          5    374.959992          -46.869999      99.9
180517  Cleats  High Value-Small Number  San Diego  EE.UU.  Mary  Allen  Corporate  CA  ...  0.13   114288  59.990002          0.48          5    260.959992          125.260002      59.9
180518  Cleats  High Value-Small Number  Sacramento  EE.UU.  Mary  Jackson  Corporate  CA  ...  0.17   60174  59.990002          0.47          5    248.960007          117.010002      59.9

180519 rows x 46 columns

In [17]: df1.groupby('Delivery Status')['order date (DateOrders)'].resample('W').count().loc['Late']

.....
Traceback (most recent call last)
Input [In [17], in ]
----> 1 df1.groupby('Delivery Status')['order date (DateOrders)'].resample('W').count().loc['Late']

File ~\anaconda3\lib\site-packages\pandas\core\groupby\groupby.py:1437, in GroupBy.resample(self, rule, *args, **kwargs)
2319 """
2320 Provide resampling when using a TimeGrouper.
2321
2322 (...)
2413 """
2414
2415 from pandas.core.resample import get_resampler_for_grouping
-> 2417 return get_resampler_for_grouping(self, rule, *args, **kwargs)

File ~\anaconda3\lib\site-packages\pandas\core\resample.py:1437, in get_resampler_for_grouping(groupby, rule, how, fill_method, limit, kind, on, **kwargs)
1435 """
1436 resample uses (key, TimeDeltaIndex).
1437 Ig TimeGrouper(freq=rule, key=on, **kwargs)
-> 1437 resampler = G.get_resampler(groupby.obj, kind=kind)
1438 return resampler._get_resampler_for_grouping(groupby)

File ~\anaconda3\lib\site-packages\pandas\core\resample.py:1599, in TimeGrouper._get_resampler(self, obj, kind)
1597 """
1598 return TimeDeltaIndexResampler(obj, groupby=self, axis=self.axis)
-> 1599 raise TypeError("Only valid with DatetimeIndex, "
1600 "TimeDeltaIndex or PeriodIndex, "
1601 "but got an instance of '%(type(ax).__name__)'"
1602 )
1603

TypeError: Only valid with DatetimeIndex, TimeDeltaIndex or PeriodIndex, but got an instance of 'RangeIndex'

In [15]: df1['order date (DateOrders)']=pd.to_datetime(df1['order date (DateOrders)'])

In [20]: df1['order date (DateOrders)']

Out[20]:
0      2017-02-26 17:17:00
1      2015-02-03 00:47:00
2      2017-01-29 04:43:00
3      2015-01-18 04:42:00
4      2017-03-02 17:36:00
...
180514  2016-12-29 06:20:00
180515  2016-03-05 07:16:00
180516  2015-12-22 10:32:00
180517  2016-10-04 11:09:00
180518  2015-12-24 01:26:00
Name: order date (DateOrders), Length: 180519, dtype: datetime64[ns]

In [1]: import pandas as pd

In [2]: import numpy as np

In [3]: from sklearn.cluster import KMeans

In [ ]: import sklearn.preprocessing import MinMaxScaler

In [6]: from matplotlib import pyplot as plt

In [7]: plt.matplotlib inline

df1=pd.read_csv('Inventory_Stock_Data.csv')

In [9]: df=pd.read_csv('Inventory_Stock_Data.csv')

In [10]: df.head()

Out[10]:
   order-now  product id  product name  avg lead time  avg order qty  current stock  max lead time  max order qty  mod  reorder point  safety stock
0         green          19  Nike Men's Fingtrap Max Training Shoe          10           1           69           30           1           1           35           25
1         orange          24  Elevation Training Mask 2.0          10           3           175           30           5           0           175           144
2         green          35  adidas Brazuca 2014 Official Match Ball          10           1           65           30           1           1           35           25
3         green          37  adidas Kids' F5 Messi FG Soccer Cleat          10           3           328           30           5           1           175           146
4         orange          44  adidas Mens F10 Messi TRX FG Soccer Cleat          10           3           175           30           5           0           175           145

In [13]: plt.scatter(df['avg lead time'],df['current stock'],df['max lead time'])

Out[13]:
<matplotlib.collections.PathCollection at 0x1f9475cb56>

In [14]: plt.scatter(df['avg order qty'],df['max order qty'],df['mod'],df['safety stock'])

Out[14]:
<matplotlib.collections.PathCollection at 0x1f9475cb56>

In [15]: df['product name'].mode()

Out[15]:
0      Adult dog supplies
1      Baby sweater
2      Bag Boy Beverage Holder
3      Bag Boy M330 Push Cart
4      Bowflex SelectTech 1090 Dumbbells
...
113  adidas Mens' F5 Messi TRX FG Soccer Cleat
114  adidas Men's F10 Messi TRX FG Soccer Cleat
115  adidas Men's Germany Black Crest Away Tee
116  adidas Youth Germany Black/Red Away Match Soc
117  insta-bed NeverFlat Air Mattress
Name: product name, Length: 118, dtype: object

In [16]: import pandas as pd

In [17]: import numpy as np

In [18]: df1=pd.read_csv('Sales_Shipment_Data.csv')

In [21]: df1.head()

Out[21]:
   Product Category Name  Class  Customer City  Customer Country  Customer Fname  Customer Lname  Customer Segment  Customer State  Order Item Discount Rate  Order Item Id  Order Item Product Price  Order Item Profit Ratio  Order Item Quantity  Order Item Total  Order Profit Per Order  Product Price  Sales
0           Women's Apparel  Moderate Value-Moderate Number  Roswell  EE.UU.  Jean  Dunn  Corporate  NM  ...  0.15    134920          50.0          0.11          2      85.0  9.610000          50.0  100.0
1           Women's Apparel  Moderate Value-Moderate Number  Los Angeles  EE.UU.  Rose  White  Corporate  CA  ...  0.16    9722          50.0          0.11          2      84.0  9.490000          50.0  100.0
2           Women's Apparel  Moderate Value-Moderate Number  Harvey  EE.UU.  Joshua  Brady  Corporate  IL  ...  0.06   129954          50.0          0.33          2      84.5  30.709999          50.0  100.0
3           Women's Apparel  Moderate Value-Moderate Number  Elmhurst  EE.UU.  Mary  Smith  Corporate  IL  ...  0.05   2929          50.0          0.29          2      95.0  27.549999          50.0  100.0
4           Women's Apparel  Moderate Value-Moderate Number  North Las Vegas  EE.UU.  Samantha  Kline  Corporate  NV  ...  0.05   140056          50.0          0.50          2      85.0  47.500000          50.0  100.0

5 rows x 46 columns

In [30]: df1['Product Name'].value_counts().sort_values()[::-1][1:10]

Out[30]:
Perfect Fitness Perfect Rip Deck          24515
Nike Men's G3 Elite 2 TD Football Cleat    22246
Nike Men's G1-FIT Victory Golf Polo        21035
O'Brien Men's Neoprene Life Vest          19298
Field & Stream Sportsman 16 Gun Fire Safe  17325
Pelican Sunstream 100 Kayak               15580
Diamondback Women's Serene Classic Comfort B1  13729
Nike Men's Free 5.0+ Running Shoe         12169
Under Armour Girls' Toddler Spine Surge Runni  19617
Fighting video games                     838
Name: Product Name, dtype: int64

In [29]: df1['Category Name'].value_counts().sort_values()[::-1][1:10]

Out[29]:
Men's Footwear          24551
Women's Apparel         22246
Indoor/Outdoor Games   21035
Fishing                 19298
Water Sports           17325
Camping & Hiking       15580
Cardio Equipment       12487
Shop by Sport          10984
Electronics            3156
Name: Category Name, dtype: int64

In [28]: df1['Customer City'].value_counts().sort_values()[::-1][1:10]

Out[28]:
Caguas          66779
Chicago         3885
Los Angeles     3417
Brooklyn        3412
New York        1616
Philadelphia    1577
Bronx           1500
San Diego       1437
Miami           1314
Houston         1297
Name: Customer City, dtype: int64

In [2]: import pandas as pd

import numpy as np

In [3]: df1=pd.read_csv('Sales_Shipment_Data.csv')

In [5]: df1['Shipping Mode'].value_counts().idxmax()

Out[5]: 'Standard Class'

In [ ]:

In [4]: import pandas as pd

In [6]: import numpy as np

df1=pd.read_csv('Sales_Shipment_Data.csv')

In [7]: df1=pd.read_csv('Sales_Shipment_Data.csv')

In [8]: df1['Category Name'].value_counts().sort_values()[::-1][1:5]

Out[8]:
Cleats          24551
Men's Footwear  22246
Women's Apparel 21035
Indoor/Outdoor Games 19298
Fishing         17325
Name: Category Name, dtype: int64

In [ ]: conda install -c conda-forge nbconvert-webpdf

In [ ]:
```