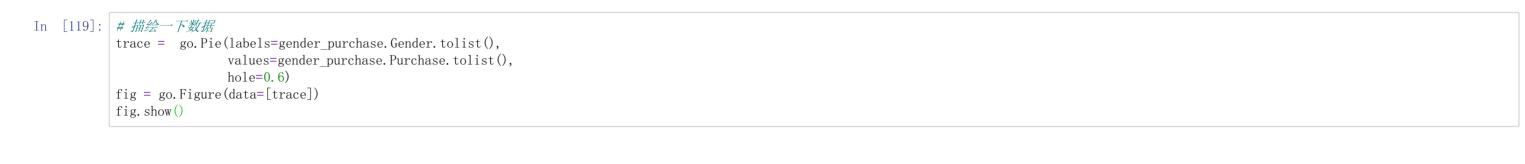
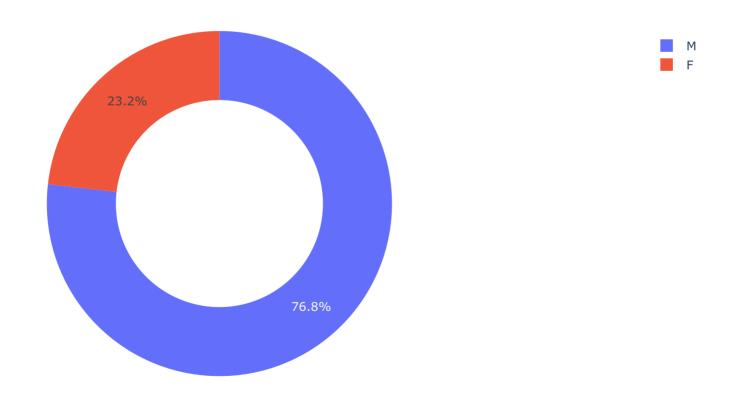
```
In [105]: | import pandas as pd
           import numpy as np
           import plotly.graph_objects as go
           import colorlover
           colors = ['#F1948A', '#AED6F1', '#F9E79F', '#E5E8E8', '#F1948A', '#D0ECE7', '#F6DDCC', '#D2B4DE',
                    '#117A65', '#FAE5D3', '#34495E', '#DC7633', '#D35400', '#0E6251', '#FCF3CF', '#E8F8F5', '#D4E6F1', '#FADBD8', '#E59866']
In [106]: f = "BlackFriday.csv"
           data = pd. read csv(f)
           data. head()
Out[106]:
               User_ID Product_ID Gender Age Occupation City_Category Stay_In_Current_City_Years Marital_Status Product_Category_1 Product_Category_2 Product_Category_3 Purchase
                       P00069042
            0 1000001
                                       F 0-17
                                                      10
                                                                                                                                             NaN
                                                                                                                                                               NaN
                                                                                                                                                                        8370
                       P00248942
                                       F 0-17
                                                      10
                                                                                             2
                                                                                                          0
                                                                                                                                              6.0
                                                                                                                                                                        15200
            1 1000001
                                                                                                                                                               14.0
            2 1000001
                       P00087842
                                       F 0-17
                                                      10
                                                                                             2
                                                                                                          0
                                                                                                                            12
                                                                                                                                             NaN
                                                                                                                                                               NaN
                                                                                                                                                                        1422
                                                                                             2
            3 1000001 P00085442
                                                                                                          0
                                                                                                                            12
                                                                                                                                                                        1057
                                       F 0-17
                                                      10
                                                                                                                                             14.0
                                                                                                                                                               NaN
                                                                    Α
            4 1000002 P00285442
                                                                                                           0
                                                                                                                                             NaN
                                                                                                                                                               NaN
                                                                                                                                                                        7969
In [107]: data. shape #显示数据大小
Out[107]: (537577, 12)
In [108]: data. Occupation. unique() #unique将数据中职业类型全部归类显示出来
Out[108]: array([10, 16, 15, 7, 20, 9, 1, 12, 17, 0, 3, 4, 11, 8, 19, 2, 18,
                  5, 14, 13, 6], dtype=int64)
In [109]: len(data. User ID. unique()) #统计人数
Out[109]: 5891
In [110]: data.describe()
Out[110]:
                                Occupation Marital_Status Product_Category_1 Product_Category_2 Product_Category_3
                      User_ID
                                                                                                                    Purchase
            count 5.375770e+05 537577.00000 537577.000000
                                                              537577.000000
                                                                                370591.000000
                                                                                                   164278.000000 537577.000000
                 1.002992e+06
                                                                  5.295546
                                                                                     9.842144
                                                                                                                  9333.859853
                                   8.08271
                                                0.408797
                                                                                                       12.669840
            mean
              std 1.714393e+03
                                   6.52412
                                                0.491612
                                                                  3.750701
                                                                                     5.087259
                                                                                                       4.124341
                                                                                                                  4981.022133
                 1.000001e+06
                                                0.000000
                                                                  1.000000
                                                                                     2.000000
                                                                                                       3.000000
                                   0.00000
                                                                                                                   185.000000
             25% 1.001495e+06
                                   2.00000
                                                0.000000
                                                                  1.000000
                                                                                     5.000000
                                                                                                       9.000000
                                                                                                                  5866.000000
                                                0.000000
                                                                  5.000000
                                                                                     9.000000
             50% 1.003031e+06
                                   7.00000
                                                                                                       14.000000
                                                                                                                  8062.000000
             75% 1.004417e+06
                                   14.00000
                                                1.000000
                                                                  8.000000
                                                                                    15.000000
                                                                                                       16.000000
                                                                                                                 12073.000000
             max 1.006040e+06
                                   20.00000
                                                1.000000
                                                                  18.000000
                                                                                    18.000000
                                                                                                      18.000000
                                                                                                                 23961.000000
In [111]: data.isna().sum()#缺失值查看
Out[111]: User_ID
                                             0
           Product_ID
                                             0
           Gender
                                             0
           Age
           Occupation
           City_Category
           Stay_In_Current_City_Years
           Marital_Status
           Product_Category_1
                                             0
           Product_Category_2
                                         166986
           Product_Category_3
                                         373299
                                             0
           Purchase
           dtype: int64
In [112]: #pivot_table 是Pandas的高级应用中的透视表的功能
            gender_purchase = data.pivot_table(values="Purchase", index=["User_ID"])
           gender_purchase. head()
Out[112]:
                        Purchase
            User_ID
                     9808.264706
            1000001
            1000002 10662.539474
            1000003 11780.517241
            1000004 15845.153846
            1000005 7745.292453
In [113]: #aggfunc指定一个函数操作,对values求和但是按照index分类
            gender_purchase = data.pivot_table(values="Purchase", aggfunc="sum", index=["User ID"])
           gender_purchase. head()
Out[113]:
                    Purchase
            User_ID
            1000001
                      333481
                      810353
            1000002
            1000003
                      341635
            1000004
                      205987
            1000005
                      821001
In [114]: #index为性别
           gender_purchase = data.pivot_table(values="Purchase",
                                              aggfunc="sum", index=["Gender"])
           gender_purchase. head()
Out[114]:
                     Purchase
            Gender
                 F 1164624021
```

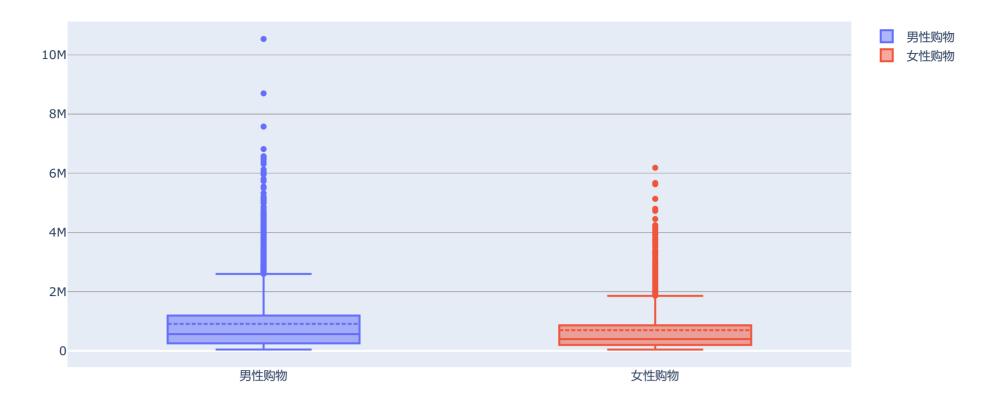
M 3853044357

```
In [115]: #index添加显示性别信息
          gender_purchase = data.pivot_table(values="Purchase",
                                          aggfunc="sum",
                                          index=["User_ID", "Gender"])
          gender_purchase.head()
Out[115]:
                          Purchase
           User_ID Gender
           1000001
                           333481
           1000002
                       M
                            810353
           1000003
                            341635
                       Μ
           1000004
                            205987
           1000005
                           821001
          上面的index不正常,需要用reset_index()更正
In [116]: #index添加显示性别信息
          gender_purchase = data.pivot_table(values="Purchase",
                                          aggfunc="sum",
                                          index=["User_ID", "Gender"]).reset_index()
                                        #reset_index 为了变为正常的index
          gender_purchase.head()
Out[116]:
             User_ID Gender Purchase
           0 1000001
                              333481
           1 1000002
                              810353
           2 1000003
                              341635
           3 1000004
                              205987
           4 1000005
                              821001
In [117]: gender_purchase.count() #count信息个数
Out[117]: User_ID
                     5891
                     5891
          Gender
                     5891
          Purchase
          dtype: int64
In [118]: gender_count = gender_purchase.groupby(by="Gender").size().reset_index(name="人数")
          print(gender_purchase.groupby(by="Gender").size())
          gender_count["占比"] = gender_count["人数"] / gender_count["人数"].sum()
          gender_count
          Gender
              1666
              4225
          dtype: int64
Out[118]:
             Gender 人数
                             占比
                  F 1666 0.282804
                  M 4225 0.717196
```

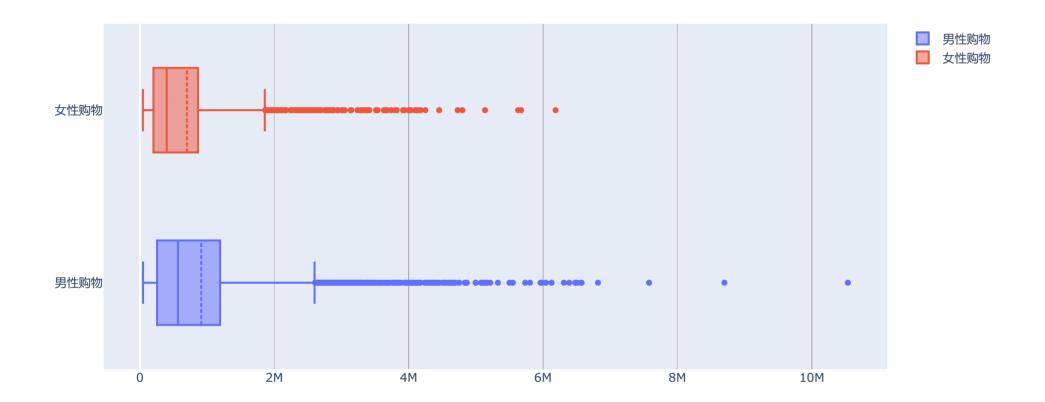




```
In [120]: y_female = gender_purchase[gender_purchase.Gender == "F"].Purchase y_male = gender_purchase[gender_purchase.Gender == "M"].Purchase trace1 = go. Box(y=y_male, name="男性购物", boxmean=True) #显示平均值(虚线) trace2 = go. Box(y=y_female, name="女性购物", boxmean=True) fig = go.Figure(data=[trace1, trace2]) fig. show()
```



```
In [121]: #機着画
y_female = gender_purchase[gender_purchase.Gender == "F"].Purchase
y_male = gender_purchase[gender_purchase.Gender == "M"].Purchase
trace1 = go. Box(x=y_male, name="男性购物", boxmean=True) #显示平均值(虚线)
trace2 = go. Box(x=y_female, name="女性购物", boxmean=True)
fig = go.Figure(data=[trace1, trace2])
fig. show()
```



Out[122]:

In [123]: #都哪些人购买了这些热销商品

top\_seller\_buyers = data[data.Product\_ID.isin(top10\_sellers.Product\_ID.tolist())]
top\_seller\_buyers.head(10)

Out[123]:

P00184942	М	46-50	7							
P00255842			1	В	2	1	1	8.0	17.0	19215
1 00200072	M	26-35	12	Α	0	0	16	NaN	NaN	20961
P00112142	М	46-50	17	В	3	1	1	2.0	14.0	19072
P00010742	М	26-35	4	В	3	1	1	8.0	17.0	19352
P00237542	М	36-45	20	С	0	1	1	15.0	16.0	11370
P00110742	М	26-35	2	В	3	0	1	2.0	8.0	15824
P00010742	F	51-55	1	В	4+	1	1	8.0	17.0	7988
P00110942	М	18-25	1	Α	1	1	1	2.0	NaN	19462
P00184942	F	26-35	1	Α	1	0	1	8.0	17.0	11715
P00255842	F	18-25	4	В	1	0	16	NaN	NaN	16416
	P00010742 P00237542 P00110742 P00010742 P00110942 P00184942	P00112142 M P00010742 M P000237542 M P000110742 M P00010742 F P00110942 M P000184942 F	P00112142 M 46-50 P00010742 M 26-35 P00237542 M 36-45 P00110742 M 26-35 P00010742 F 51-55 P00110942 M 18-25 P00184942 F 26-35	P00112142 M 46-50 17 P00010742 M 26-35 4 P000237542 M 36-45 20 P00110742 M 26-35 2 P00010742 F 51-55 1 P00110942 M 18-25 1 P00184942 F 26-35 1	B       P00112142       M       46-50       17       B         B       P00010742       M       26-35       4       B         B       P00237542       M       36-45       20       C         B       P00110742       M       26-35       2       B         B       P00010742       F       51-55       1       B         B       P00110942       M       18-25       1       A         B       P00184942       F       26-35       1       A	P00112142       M       46-50       17       B       3         P00010742       M       26-35       4       B       3         P00237542       M       36-45       20       C       0         P00110742       M       26-35       2       B       3         P00010742       F       51-55       1       B       4+         P00110942       M       18-25       1       A       1         P00184942       F       26-35       1       A       1	R P00112142       M 46-50       17       B       3       1         R P00010742       M 26-35       4       B       3       1         R P00237542       M 36-45       20       C       0       1         R P00110742       M 26-35       2       B       3       0         R P00010742       F 51-55       1       B       4+       1         R P00110942       M 18-25       1       A       1       1         R P00184942       F 26-35       1       A       1       0	P00112142       M 46-50       17       B       3       1       1         P00010742       M 26-35       4       B       3       1       1         P00237542       M 36-45       20       C       0       1       1         P00110742       M 26-35       2       B       3       0       1         P00010742       F 51-55       1       B       4+       1       1         P00110942       M 18-25       1       A       1       0       1         P00184942       F 26-35       1       A       1       0       1	4       P00112142       M       46-50       17       B       3       1       1       2.0         8       P00010742       M       26-35       4       B       3       1       1       8.0         8       P00237542       M       36-45       20       C       0       1       1       15.0         8       P00110742       M       26-35       2       B       3       0       1       2.0         9       P00010742       F       51-55       1       B       4+       1       1       1       8.0         9       P00110942       M       18-25       1       A       1       1       1       2.0         9       P00184942       F       26-35       1       A       1       0       1       8.0	4 P00112142       M 46-50       17       B       3       1       1       2.0       14.0         8 P00010742       M 26-35       4       B       3       1       1       8.0       17.0         6 P00237542       M 36-45       20       C       0       1       1       15.0       16.0         8 P00110742       M 26-35       2       B       3       0       1       2.0       8.0         9 P0010742       F 51-55       1       B       4+       1       1       8.0       17.0         8 P00110942       M 18-25       1       A       1       1       1       2.0       NaN         9 P00184942       F 26-35       1       A       1       0       1       8.0       17.0

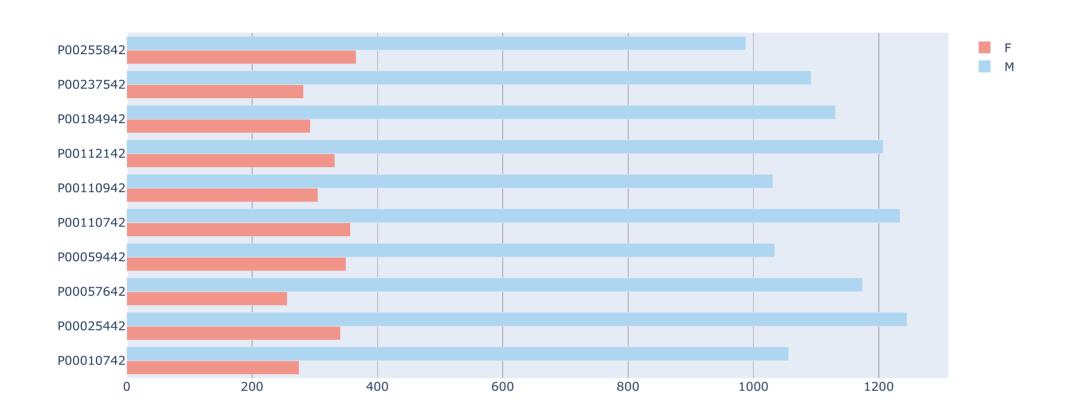
```
aggfunc="count").reset_index()
    top_seller_gender.head(10)
```

## Out[124]:

	Product_ID	Gender	Purchase
0	P00010742	F	275
1	P00010742	M	1056
2	P00025442	F	341
3	P00025442	М	1245
4	P00057642	F	256
5	P00057642	M	1174
6	P00059442	F	350
7	P00059442	М	1034
8	P00110742	F	357
9	P00110742	М	1234

```
In [125]: traces = []
```

```
i = 0
for g in top_seller_gender.Gender.unique():
        trace = go.Bar(x=top_seller_gender[top_seller_gender.Gender==g].Purchase,
                      y=top_seller_gender[top_seller_gender.Gender==g].Product_ID,
                      marker=dict(color=colors[i]),
                      orientation="h")
        traces.append(trace)
        i += 1
fig = go.Figure(data=traces)
fig.show()
```



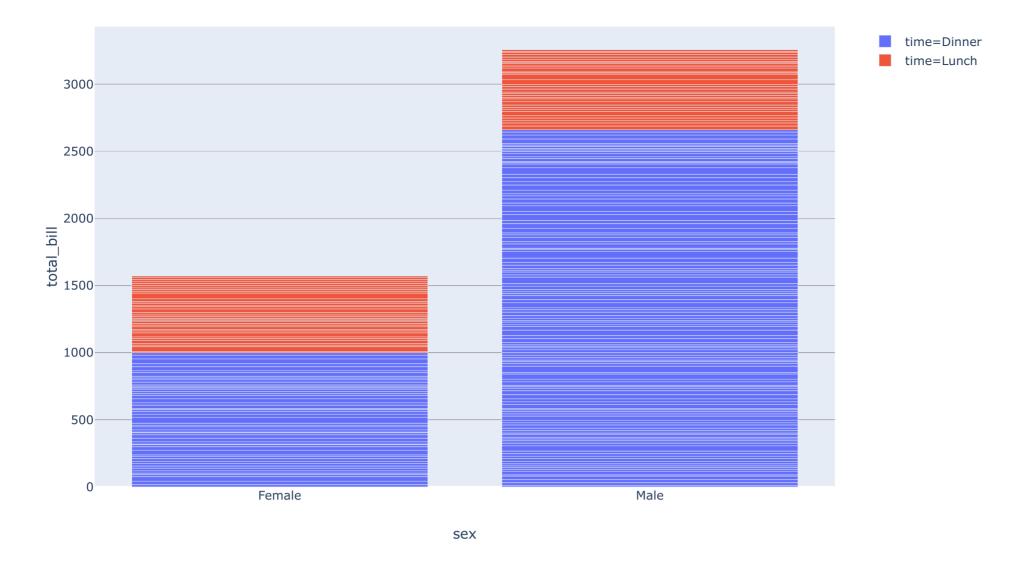
## In [126]: top\_seller\_city = top\_seller\_buyers.pivot\_table(values="Purchase", index=["Product\_ID", "City\_Category"], aggfunc="count").reset\_index()

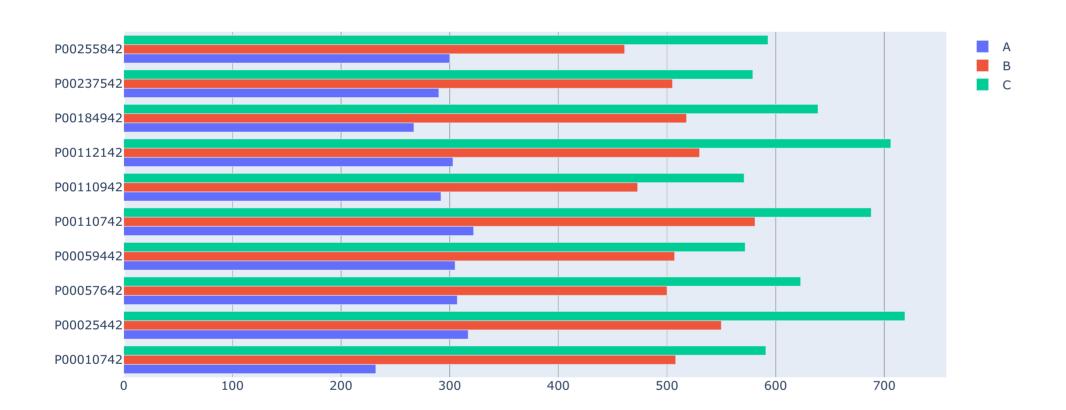
top\_seller\_city.head(10)

## Out[126]:

	Product_ID	City_Category	Purchase
0	P00010742	А	232
1	P00010742	В	508
2	P00010742	С	591
3	P00025442	Α	317
4	P00025442	В	550
5	P00025442	С	719
6	P00057642	Α	307
7	P00057642	В	500
8	P00057642	С	623
9	P00059442	Α	305

```
In [127]: import plotly.express as px
tips = px. data.tips()
fig = px.bar(tips, x="sex", y="total_bill", color="time")
fig.show()
```





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