```
In [1]: #importing Libraries
    import numpy as np
    import pandas as pd

import plotly.express as px
    import plotly.graph_objs as go

from datetime import date

In [2]: data = pd.read_csv("kz.csv")
```

About this file

This file contains purchase data from April 2020 to November 2020 from a large home appliances and electronics online store, collected by Open CDP project.

In [3]:	data.head(10)						
Out[3]:		event_time	order_id	product_id	category_id	category_code	
	0	2020-04-24 11:50:39 UTC	2294359932054536986	1515966223509089906	2.268105e+18	electronics.tablet	sa
	1	2020-04-24 11:50:39 UTC	2294359932054536986	1515966223509089906	2.268105e+18	electronics.tablet	sa
	2	2020-04-24 14:37:43 UTC	2294444024058086220	2273948319057183658	2.268105e+18	electronics.audio.headphone	ŀ
	3	2020-04-24 14:37:43 UTC	2294444024058086220	2273948319057183658	2.268105e+18	electronics.audio.headphone	ŀ
	4	2020-04-24 19:16:21 UTC	2294584263154074236	2273948316817424439	2.268105e+18	NaN	k
	5	2020-04-26 08:45:57 UTC	2295716521449619559	1515966223509261697	2.268105e+18	furniture.kitchen.table	m
	6	2020-04-26 09:33:47 UTC	2295740594749702229	1515966223509104892	2.268105e+18	electronics.smartphone	
	7	2020-04-26 09:33:47 UTC	2295740594749702229	1515966223509104892	2.268105e+18	electronics.smartphone	
	8	2020-04-26 09:33:47 UTC	2295740594749702229	1515966223509104892	2.268105e+18	electronics.smartphone	
	9	2020-04-26 09:33:47 UTC	2295740594749702229	1515966223509104892	2.268105e+18	electronics.smartphone	

```
In [4]:
          data.dtypes
 Out[4]: event_time
                            object
         order_id
                             int64
         product_id
                             int64
                           float64
         category_id
                            object
         category code
         brand
                            object
                           float64
         price
         user id
                           float64
         dtype: object
         Cleaning the Data
 In [5]:
          data.set_index('order_id', inplace = True)
 In [6]:
          null columns = data.columns[data.isnull().any()]
          data[null columns].isnull().sum()
 Out[6]: category_id
                            431954
         category_code
                            612202
         brand
                            506005
         price
                            431954
                           2069352
         user id
         dtype: int64
 In [7]:
          n_unique_products = data['product_id'].nunique()
          n unique users = data['user id'].nunique()
          print('Number of unique users: ' + str(n unique users) + '. Number of unique products i
         Number of unique users: 98262. Number of unique products is: 25113
 In [8]:
          data['event_time'] = pd.to_datetime(data['event_time'])
In [10]:
          data.dropna(subset = ['category_code'], inplace = True)
          # Getting the main category from the category_code
In [11]:
          data['category'] = data['category_code'].str.rsplit('.', n=1, expand = True)[1]
          data.drop(columns = ['category code'], inplace = True)
```

Analyzing data

Best performing brands

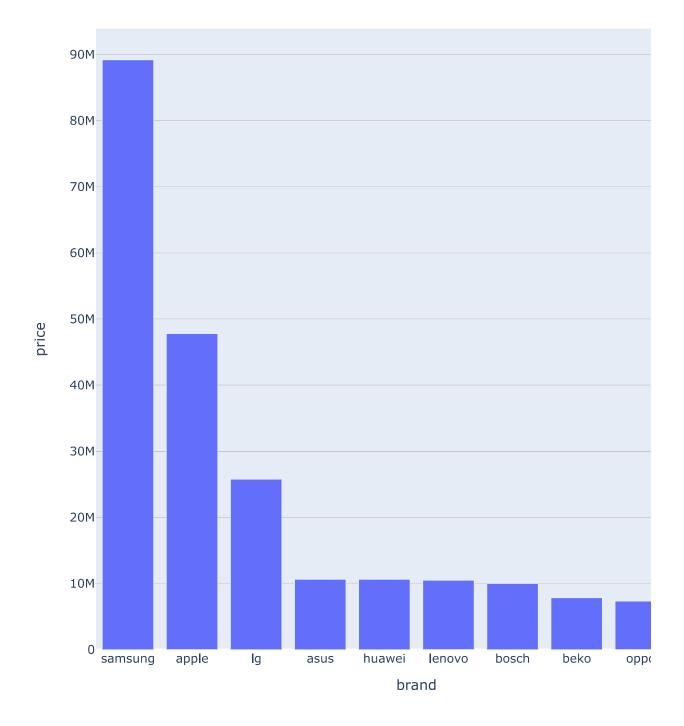
```
In [12]: # best performing brands

best_performing_brands = data.groupby('brand')['price'].sum().reset_index().sort_values

In [13]: fig = px.bar(
    best_performing_brands,
    x = 'brand',
    y = 'price',
    title = 'Best performing brands',
    width = 800,
    height = 800
    )
```

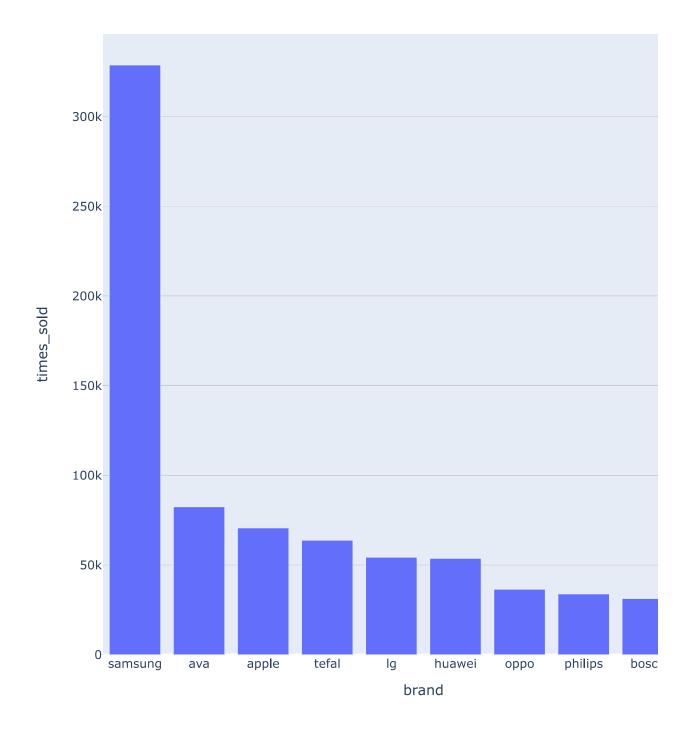
fig.show()

Best performing brands



```
In [14]: #most sold brands
most_sold_brands = data.groupby('brand')['price'].agg('count').reset_index().sort_value
most_sold_brands.rename(columns = {"brand":"brand", "price": "times_sold"}, inplace = T
```

Most sold brands

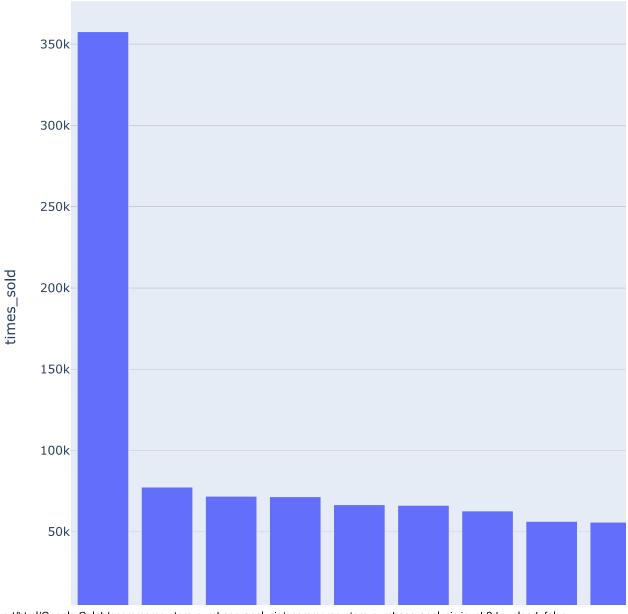


```
In [18]: #most bought categories
   most_sold_categories = data.groupby('category')['price'].agg('count').reset_index().sor
   most_sold_categories.rename(columns = {"category": 'category', "price": "times_sold"},

In [21]: fig = px.bar(
   most_sold_categories,
        x = 'category',
        y = 'times_sold',
        title = 'Most sold categories',
        width = 800,
        height = 800
)

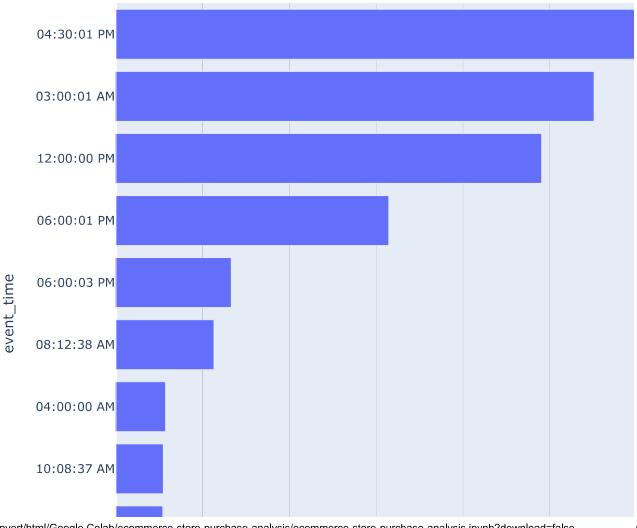
fig.show()
```

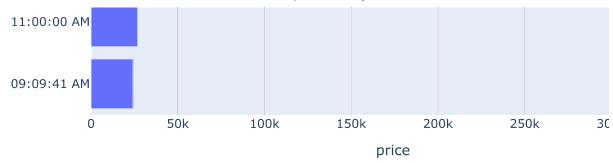
Most sold categories



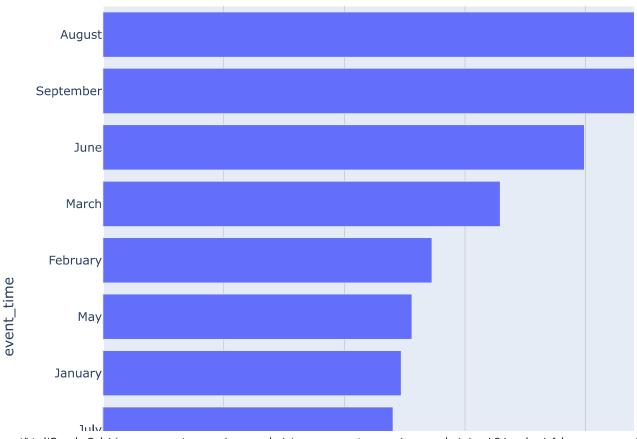


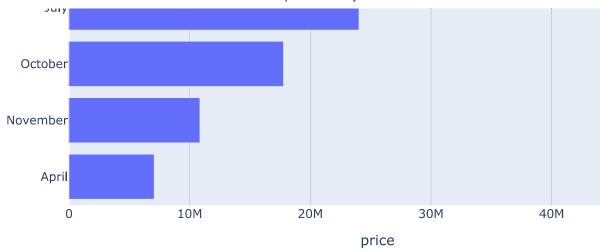
At what time most of the purchases were made





In which months most of the purchases were made





```
In [28]: # How much money spent 20% of top buyers in comparison with other 80% clients
    most_active_users = data.groupby('user_id')['price'].sum().reset_index().sort_values('price'].sum().reset_index().sort_values('top_20_percent_buyers = most_active_users['price'].sum()
    bottom_80_percent_buyers = least_active_users['price'].sum()
    last_data = pd.DataFrame(data = {'most_active': [most_active_users['price'].sum()], 'le

In [29]: fig = px.bar(
    last_data,
    title = "20% of most active in comparison with other 80% of buyers",
    width = 800,
    height = 800
)

fig.show()
```

20% of most active in comparison with other 80% of buyers



