Title: Analyzing video games sales ¶

Information about the data used:

- -Data Source: Video Game Sales Dataset
- -Location: https://www.kaggle.com/gregorut/videogamesales

(https://www.kaggle.com/gregorut/videogamesales)

-Filename: videogamesales.zip

Initial the exploration of the Dataset

```
In [5]: import numpy as np
import pandas as pd
```

Importing data

```
df = pd.read_csv('vgsales.csv')
 In [6]:
 In [7]:
           df.head()
 Out[7]:
              Rank
                            Name Platform
                                             Year
                                                     Genre
                                                            Publisher NA_Sales EU_Sales JP_Sales Other
            0
                  1
                        Wii Sports
                                       Wii
                                           2006.0
                                                     Sports
                                                             Nintendo
                                                                          41.49
                                                                                    29.02
                                                                                               3.77
                       Super Mario
            1
                  2
                                      NES
                                           1985.0 Platform
                                                             Nintendo
                                                                          29.08
                                                                                     3.58
                                                                                               6.81
                            Bros.
            2
                     Mario Kart Wii
                                           2008.0
                                                             Nintendo
                                       Wii
                                                    Racing
                                                                          15.85
                                                                                    12.88
                                                                                               3.79
                        Wii Sports
            3
                  4
                                       Wii 2009.0
                                                     Sports
                                                             Nintendo
                                                                          15.75
                                                                                    11.01
                                                                                               3.28
                           Resort
                         Pokemon
                                                      Role-
                  5 Red/Pokemon
                                       GB 1996.0
                                                             Nintendo
                                                                          11.27
                                                                                     8.89
                                                                                              10.22
                                                    Playing
                             Blue
 In [9]: | df.shape
 Out[9]: (16598, 11)
In [10]: | df.columns
Out[10]: Index(['Rank', 'Name', 'Platform', 'Year', 'Genre', 'Publisher', 'NA_Sales',
                    'EU_Sales', 'JP_Sales', 'Other_Sales', 'Global_Sales'],
                  dtype='object')
```

Out[12]:								
		Rank	Year	NA_Sales	EU_Sales	JP_Sales	Other_Sales	Globa
	count	16598.000000	16327.000000	16598.000000	16598.000000	16598.000000	16598.000000	16598
	mean	8300.605254	2006.406443	0.264667	0.146652	0.077782	0.048063	0
	std	4791.853933	5.828981	0.816683	0.505351	0.309291	0.188588	1
	min	1.000000	1980.000000	0.000000	0.000000	0.000000	0.000000	0
	25%	4151.250000	2003.000000	0.000000	0.000000	0.000000	0.000000	0
	50%	8300.500000	2007.000000	0.080000	0.020000	0.000000	0.010000	0
	75%	12449.750000	2010.000000	0.240000	0.110000	0.040000	0.040000	0
	max	16600.000000	2020.000000	41.490000	29.020000	10.220000	10.570000	82
	4							•

Know we will clean our data but let us first check if there is any null value

```
In [39]: df.isnull().values.any()
Out[39]: True
In [40]: clean_df= df.dropna()
In [15]: clean_df.isnull().values.any()
Out[15]: False
```

Dimentions of new cleaned data

In [12]: df.describe()

```
In [18]: clean_df.shape
Out[18]: (16291, 11)
```

```
In [19]: clean_df.info()
         <class 'pandas.core.frame.DataFrame'>
         Int64Index: 16291 entries, 0 to 16597
         Data columns (total 11 columns):
         Rank
                          16291 non-null int64
                          16291 non-null object
         Name
         Platform
                          16291 non-null object
                          16291 non-null float64
         Year
         Genre
                          16291 non-null object
         Publisher
                          16291 non-null object
         NA Sales
                          16291 non-null float64
         EU Sales
                          16291 non-null float64
         JP Sales
                          16291 non-null float64
         Other Sales
                          16291 non-null float64
                         16291 non-null float64
         Global_Sales
         dtypes: float64(6), int64(1), object(4)
         memory usage: 1.5+ MB
```

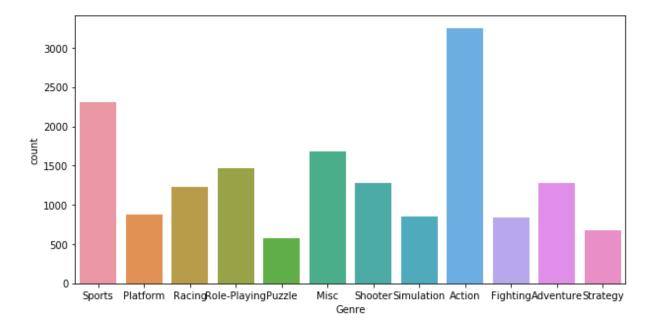
Importing seaborn and matplotlib

```
In [22]: import matplotlib.pyplot as plt
   import seaborn as sns
   %matplotlib inline
```

Data Visulation

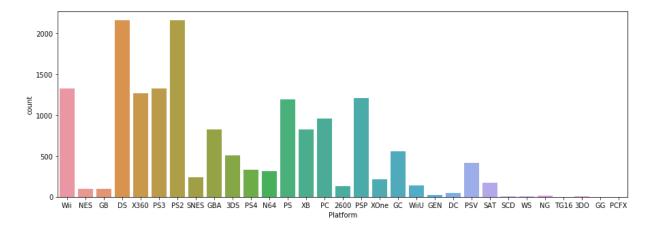
```
In [43]: plt.figure(figsize=(10,5))
    sns.countplot(clean_df['Genre'])
```

Out[43]: <matplotlib.axes._subplots.AxesSubplot at 0x24b4a8ba5c0>



```
In [44]: plt.figure(figsize=(15,5))
    sns.countplot(df['Platform'])
```

Out[44]: <matplotlib.axes._subplots.AxesSubplot at 0x24b49d09400>

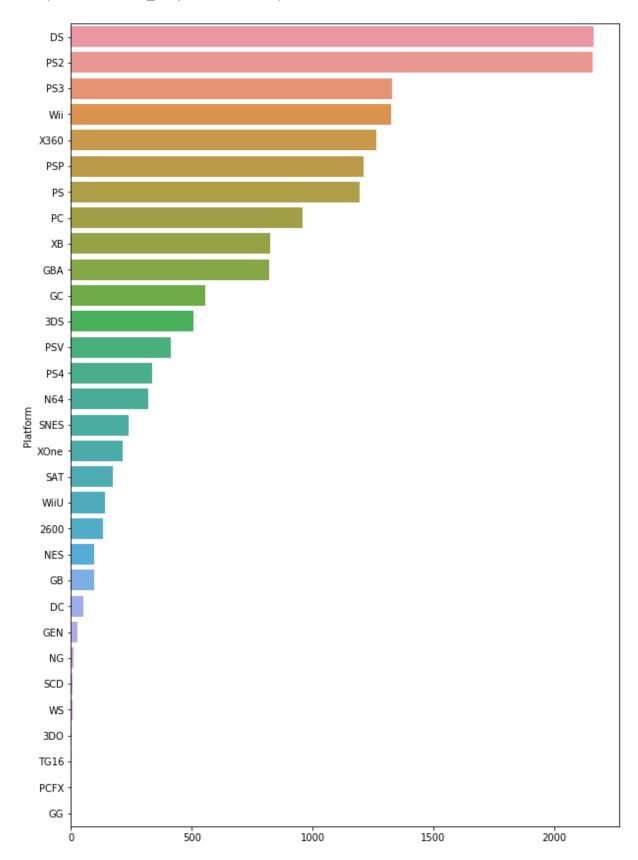


Out[45]:

Genre	Action	Adventure	Fighting	Misc	Platform	Puzzle	Racing	Role- Playing	Shooter	Simulatio
Platform										
2600	61	2	2	5	9	11	6	0	24	_
3DO	0	1	0	0	0	1	0	0	0	
3DS	182	37	14	53	28	20	11	86	7	3
DC	3	11	12	0	2	0	6	4	3	
DS	343	240	36	393	92	238	67	200	42	28
4										

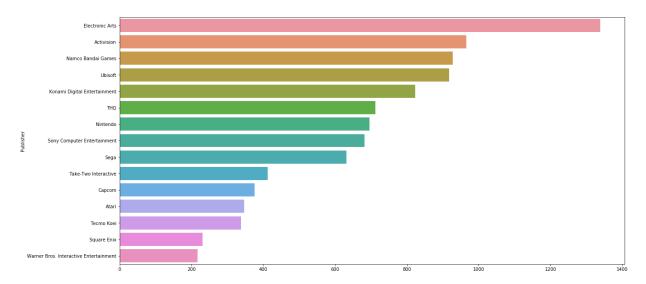
```
In [46]: platGenreTotal =platGenre.sum(axis=1).sort_values(ascending = False)
    plt.figure(figsize=(10,15))
    sns.barplot(x=platGenreTotal.values,y=platGenreTotal.index)
```

Out[46]: <matplotlib.axes._subplots.AxesSubplot at 0x24b49722cc0>



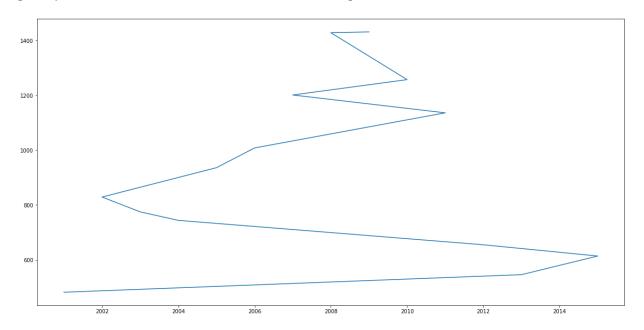
So You can see DS and PS2 have the most games in their platform and now, we can see the detail of genre of game in platform that have more than 1000 games. In the seaborn, it is easy to use heatrmap rather than the stacked bar, so we can use the heatmap to have a try.

Out[47]: <matplotlib.axes._subplots.AxesSubplot at 0x24b461e8278>



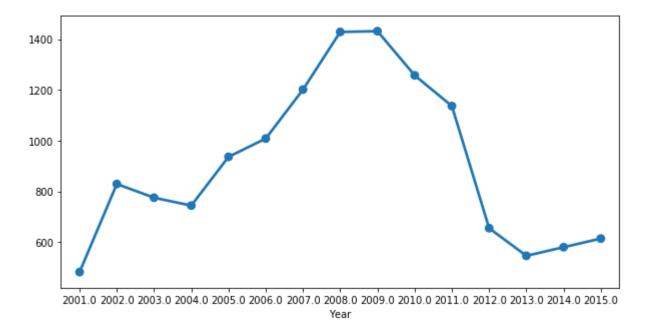
In [49]: yr=clean_df.groupby('Year')['Year'].count().sort_values(ascending = False).head()
plt.plot(yr)

Out[49]: [<matplotlib.lines.Line2D at 0x24b49c657b8>]



```
In [50]: plt.figure(figsize=(10,5))
    sns.pointplot(x=yr.index ,y=yr.values)
```

Out[50]: <matplotlib.axes._subplots.AxesSubplot at 0x24b49c43f60>



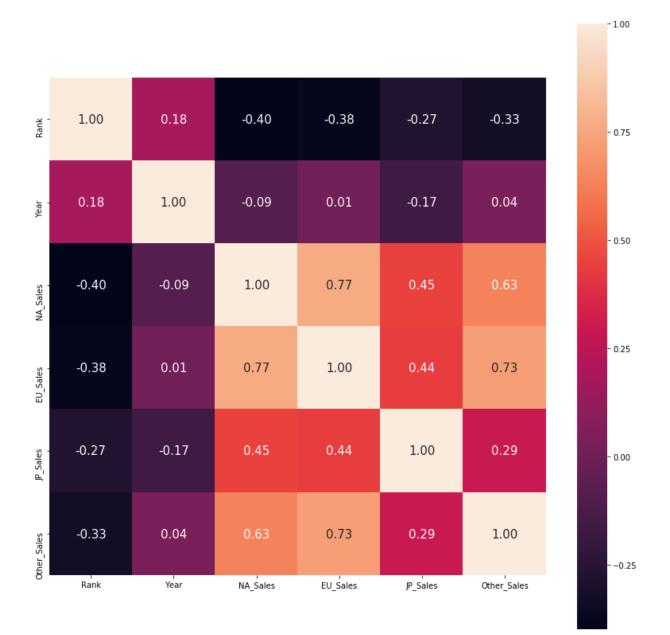
So it is showing most important year is 2009 based on others year convert Categorical variable to numerical Variables

```
In [73]: from sklearn import model selection, preprocessing
         for c in clean df.columns:
              if clean df[c].dtype == 'object':
                  lbl = preprocessing.LabelEncoder()
                  lbl.fit(list(clean df[c].values))
                  df[c] = lbl.transform(list(clean_df[c].values))
         ValueError
                                                    Traceback (most recent call last)
         <ipython-input-73-73b8d0b09c90> in <module>
                         lbl = preprocessing.LabelEncoder()
               5
                         lbl.fit(list(clean df[c].values))
                         df[c] = lbl.transform(list(clean df[c].values))
         ---> 6
               7
         ~\Anaconda3\lib\site-packages\pandas\core\frame.py in __setitem__(self, key, va
         lue)
            3117
                         else:
            3118
                             # set column
         -> 3119
                              self._set_item(key, value)
            3120
                     def _setitem_slice(self, key, value):
            3121
         ~\Anaconda3\lib\site-packages\pandas\core\frame.py in set item(self, key, valu
         e)
            3192
            3193
                         self. ensure valid index(value)
                         value = self._sanitize_column(key, value)
         -> 3194
                         NDFrame._set_item(self, key, value)
            3195
            3196
         ~\Anaconda3\lib\site-packages\pandas\core\frame.py in _sanitize_column(self, ke
         y, value, broadcast)
            3389
            3390
                             # turn me into an ndarray
                             value = sanitize index(value, self.index, copy=False)
         -> 3391
                              if not isinstance(value, (np.ndarray, Index)):
            3392
            3393
                                  if isinstance(value, list) and len(value) > 0:
         ~\Anaconda3\lib\site-packages\pandas\core\series.py in sanitize index(data, in
         dex, copy)
            3999
                     if len(data) != len(index):
            4000
         -> 4001
                         raise ValueError('Length of values does not match length of '
          'index')
            4002
            4003
                     if isinstance(data, ABCIndexClass) and not copy:
```

ValueError: Length of values does not match length of index

```
AttributeError Traceback (most recent call last)
<ipython-input-74-8c8a82c585d2> in <module>
7 xticklabels=corr.columns.values,
8 yticklabels=corr.columns.values)
----> 9 sns.plt.title('Heatmap of Correlation Matrix')
```

AttributeError: module 'seaborn' has no attribute 'plt'



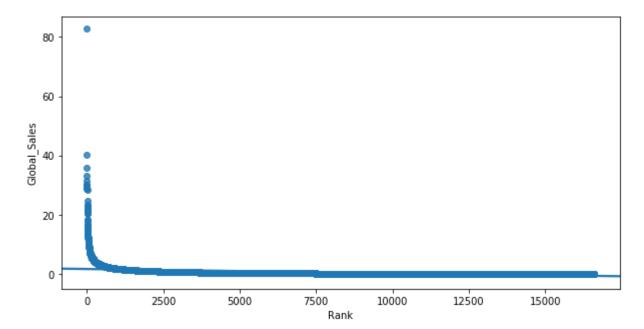
so all sales are highly correlated with global sales which is very true..Rank is not showing highly correlated with global sales as this is shown in Matrix formation.

```
In [53]: plt.figure(figsize=(10,5))
    sns.regplot(x='Rank',y='Global_Sales',data=clean_df)
```

C:\Users\Aymen\Anaconda3\lib\site-packages\scipy\stats\stats.py:1713: FutureWar ning: Using a non-tuple sequence for multidimensional indexing is deprecated; u se `arr[tuple(seq)]` instead of `arr[seq]`. In the future this will be interpre ted as an array index, `arr[np.array(seq)]`, which will result either in an err or or a different result.

return np.add.reduce(sorted[indexer] * weights, axis=axis) / sumval

Out[53]: <matplotlib.axes._subplots.AxesSubplot at 0x24b4cc18978>



```
In [55]: plt.figure(figsize=(10,5))
          sns.regplot(x='NA_Sales',y='Global_Sales',data=clean_df)
Out[55]: <matplotlib.axes._subplots.AxesSubplot at 0x24b4d507278>
             80
             60
          Global_Sales
             20
                                   10
                                                    20
                                                                     30
                                                                                     40
                                                   NA Sales
In [56]:
         clean_df.dtypes
Out[56]: Rank
                             int64
          Name
                            object
          Platform
                           object
                          float64
          Year
                           object
          Genre
          Publisher
                           object
          NA_Sales
                          float64
          EU Sales
                          float64
          JP Sales
                          float64
          Other_Sales
                          float64
          Global Sales
                          float64
          dtype: object
         #Train-Test split
In [57]:
          from sklearn.model_selection import train_test_split
          label = clean_df.pop('Global_Sales')
          data_train, data_test, label_train, label_test = train_test_split(clean_df, label_
```

```
In [58]: data_train.shape,data_test.shape
```

Out[58]: ((13032, 10), (3259, 10))

Start with Extream gradient boosting

```
In [65]: #!pip install xqboost
         Collecting xgboost
           Downloading https://files.pythonhosted.org/packages/51/c1/198915b13e98b62a98f
         48309c41012638464651da755d941f4abe384c012/xgboost-0.82-py2.py3-none-win_amd64.w
         hl (https://files.pythonhosted.org/packages/51/c1/198915b13e98b62a98f48309c4101
         2638464651da755d941f4abe384c012/xgboost-0.82-py2.py3-none-win amd64.whl) (7.7M
         B)
         Requirement already satisfied: scipy in c:\users\aymen\anaconda3\lib\site-packa
         ges (from xgboost) (1.1.0)
         Requirement already satisfied: numpy in c:\users\aymen\anaconda3\lib\site-packa
         ges (from xgboost) (1.15.4)
         Installing collected packages: xgboost
         Successfully installed xgboost-0.82
In [66]: import xgboost as xgb
         from sklearn.model selection import KFold, train test split, GridSearchCV
In [67]: | xgb_params = {
              'eta': 0.05,
              'max_depth': 5,
             'subsample': 0.7,
             'colsample bytree': 0.7,
             'objective': 'reg:linear',
             'eval_metric': 'rmse',
```

'silent': 1

```
In [69]: | dtrain = xgb.DMatrix(data train, label train)
         ValueError
                                                    Traceback (most recent call last)
         <ipython-input-69-f2c2fb056c93> in <module>
         ----> 1 dtrain = xgb.DMatrix(data_train, label_train)
         ~\Anaconda3\lib\site-packages\xgboost\core.py in init (self, data, label, mi
         ssing, weight, silent, feature_names, feature_types, nthread)
             382
                         data, feature_names, feature_types = _maybe_pandas_data(data,
             383
                                                                                  feature
         _names,
                                                                                  feature
         --> 384
         types)
             385
             386
                         data, feature_names, feature_types = _maybe_dt_data(data,
         ~\Anaconda3\lib\site-packages\xgboost\core.py in maybe pandas data(data, featu
         re names, feature types)
                         msg = """DataFrame.dtypes for data must be int, float or bool.
             239
             240
                                  Did not expect the data types in fields """
                         raise ValueError(msg + ', '.join(bad_fields))
         --> 241
             242
             243
                     if feature names is None:
         ValueError: DataFrame.dtypes for data must be int, float or bool.
                         Did not expect the data types in fields Name, Platform, Genre,
          Publisher
         lr_data_train=data_train[['NA_Sales','EU_Sales','JP_Sales','Other_Sales']]
In [70]:
         lr_data_test=data_test[['NA_Sales','EU_Sales','JP_Sales','Other_Sales']]
         lr label train=label train
         lr_label_test=label_test
In [71]: #Linear Regression
         from sklearn.linear_model import LinearRegression
         lr = LinearRegression()
         lr.fit(lr data train, lr label train)
         lr score train = lr.score(lr data train, lr label train)
         print("Training score: ",lr_score_train)
         lr score test = lr.score(lr data test, lr label test)
         print("Testing score: ",lr_score_test)
         Training score: 0.9999889271662358
         Testing score: 0.9999887509682348
```

The accuracy improved to 99%

```
In [75]: y_pre = lr.predict(lr_data_test)
```

In [76]: out_lr = pd.DataFrame({'Actual_Global_Sales': lr_label_test, 'Predict_Global_Sale
 out_lr[['Actual_Global_Sales','Predict_Global_Sales','Diff']].head(5)

Out[76]:

	Actual_Global_Sales	Predict_Global_Sales	Diff
9317	0.14	0.130333	0.009667
14835	0.03	0.020339	0.009661
9752	0.12	0.120335	-0.000335
10251	0.11	0.110324	-0.000324
16565	0.01	0.010337	-0.000337

In [77]: out_lr.shape

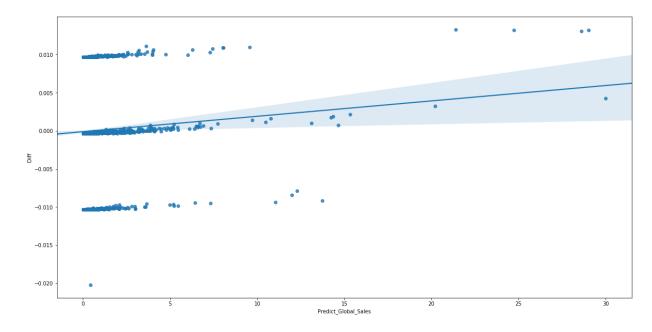
Out[77]: (3259, 3)

In [78]: sns.regplot(out_lr['Predict_Global_Sales'],out_lr['Diff'])

C:\Users\Aymen\Anaconda3\lib\site-packages\scipy\stats\stats.py:1713: FutureWar ning: Using a non-tuple sequence for multidimensional indexing is deprecated; u se `arr[tuple(seq)]` instead of `arr[seq]`. In the future this will be interpre ted as an array index, `arr[np.array(seq)]`, which will result either in an err or or a different result.

return np.add.reduce(sorted[indexer] * weights, axis=axis) / sumval

Out[78]: <matplotlib.axes. subplots.AxesSubplot at 0x24b4cac4908>



```
In [79]:
         #Ensemble XGBOOST & LINEAR REGRESSOR for train data
         en dtest=xgb.DMatrix(data train)
         y_xgb_pred = model.predict(en_dtest)
         y_lr_pred = lr.predict(lr_data_train)
         Ensemble=pd.DataFrame({'XGBOOST':y_xgb_pred ,'LINEAR_REG':y_lr_pred ,'GLOBAL_SAL
         ValueError
                                                    Traceback (most recent call last)
         <ipython-input-79-d8edccb37eeb> in <module>
               1 #Ensemble XGBOOST & LINEAR REGRESSOR for train data
         ---> 2 en dtest=xgb.DMatrix(data train)
               3 y xgb pred = model.predict(en dtest)
               5 y lr pred = lr.predict(lr data train)
         ~\Anaconda3\lib\site-packages\xgboost\core.py in __init__(self, data, label, mi
         ssing, weight, silent, feature names, feature types, nthread)
             382
                         data, feature_names, feature_types = _maybe_pandas_data(data,
             383
                                                                                  feature
         _names,
         --> 384
                                                                                  feature
         _types)
             385
             386
                         data, feature names, feature types = maybe dt data(data,
         ~\Anaconda3\lib\site-packages\xgboost\core.py in maybe pandas data(data, featu
         re names, feature types)
                         msg = """DataFrame.dtypes for data must be int, float or bool.
             239
                                  Did not expect the data types in fields """
             240
                         raise ValueError(msg + ', '.join(bad_fields))
         --> 241
             242
             243
                     if feature names is None:
         ValueError: DataFrame.dtypes for data must be int, float or bool.
                         Did not expect the data types in fields Name, Platform, Genre,
```

Publisher

```
In [80]: corr=Ensemble.corr()
         corr = (corr)
         sns.heatmap(corr,
                     xticklabels=corr.columns.values,
                     yticklabels=corr.columns.values)
         sns.plt.title('Heatmap of Correlation Matrix')
                                                    Traceback (most recent call last)
         NameError
         <ipython-input-80-35c59d31881a> in <module>
         ----> 1 corr=Ensemble.corr()
               2 corr = (corr)
               3 sns.heatmap(corr,
                             xticklabels=corr.columns.values,
               4
               5
                             yticklabels=corr.columns.values)
         NameError: name 'Ensemble' is not defined
In [ ]:
In [ ]:
In [ ]:
```