	Data Description The file train.csv contains metrics and other details of about 15000 youtube videos. The metrics include number of views, likes, dislikes, comments and apart from that published date, duration
	and category are also included. The train.csv file also contains the metric number of adviews which is our target variable for prediction Importing Libraries
In []:	<pre>import nampy as np import pandas as pd import matplotlib.cm as cm import matplotlib.pyplot as plt import warnings</pre>
	warnings.filterwarnings("ignore") Importing data
In [3]: In [4]:	data_train = pd.read_csv(path + "youtubead_train.csv")
Out[4]:	vidid adview views likes dislikes comment published duration category 0 VID_18655 40 1031602 8523 363 1095 9/14/2016 PT7M37S F
	1 VID_14135 2 1707 56 2 6 10/1/2016 PT9M30S D 2 VID_2187 1 2023 25 0 2 7/2/2016 PT2M16S C 3 VID_23096 6 620860 777 161 153 7/27/2016 PT4M22S H 4 VID_10175 1 666 1 0 0 6/29/2016 PT31S D
In [5]: Out[5]:	data_train.shape (14999, 9)
In [7]:	Assigning each category a number for Category feature
Out[7]:	<pre>data_train["category"] = data_train["category"].map(category) data_train.head()</pre>
	0 VID_18655 40 1031602 8523 363 1095 9/14/2016 PT7M37S 6 1 VID_14135 2 1707 56 2 6 10/1/2016 PT9M30S 4 2 VID_2187 1 2023 25 0 2 7/2/2016 PT2M16S 3 2 VID_2387 6 500060 777 464 152 7/2/2016 PT2M16S 3
	3 VID_23096 6 620860 777 161 153 7/27/2016 PT4M22S 8 4 VID_10175 1 666 1 0 0 6/29/2016 PT31S 4 Removing character "F" present in data
In [8]:	<pre>data_train = data_train[data_train.views!='F'] data_train = data_train[data_train.likes!='F'] data_train = data_train[data_train.dislikes!='F']</pre>
	Convert values to integers for views, likes, dislikes, comments and adview
In [9]:	<pre>data_train["views"] = pd.to_numeric(data_train["views"]) data_train["comment"] = pd.to_numeric(data_train["comment"]) data_train["likes"] = pd.to_numeric(data_train["likes"]) data_train["dislikes"] = pd.to_numeric(data_train["dislikes"]) data_train["adview"] = pd.to_numeric(data_train["adview"])</pre>
In [10]:	COIDMIN_VIGIO = Gata_train[Vigio]
In [11]:	<pre>from sklearn.preprocessing import LabelEncoder data_train['duration'] = LabelEncoder().fit_transform(data_train['duration']) data_train['vidid'] = LabelEncoder().fit_transform(data_train['vidid'])</pre>
In [12]:	<pre>data_train['published'] = LabelEncoder().fit_transform(data_train['published']) data_train.head()</pre>
Out[12]:	0 5912 40 1031602 8523 363 1095 2235 2925 6 1 2741 2 1707 56 2 6 207 3040 4
	2 8138 1 2023 25 0 2 1905 1863 3 3 9005 6 620860 777 161 153 1952 2546 8 4 122 1 666 1 0 0 1783 1963 4
In [16]:	Convert Time_in_sec for duration import time import datetime
In [17]:	<pre>def checki(x): y = x[2:] h = ''</pre>
	<pre>m = '' s = '' nm = '' p = ['H', 'M', 'S'] for i in y: if i not in p:</pre>
	<pre>nm+=i else: if(i=="H"): h = nm nm = '' elif(i=="H"):</pre>
	<pre>m = nm nm = '' else: s = nm nm = '' if(h==''):</pre>
	h = '00' if(m == ''): m = '00' if(s==''): s = '00' bp = h+':'+m+':'+s
In [20]:	<pre>train = pd.read_csv("youtubead_train.csv") mp = pd.read_csv(path + "youtubead_train.csv")["duration"]</pre>
In [23]:	<pre>time = mp.apply(checki) def func_sec(time_string): h, m, s = time_string.split(':') return int(h) * 3600 + int(m) * 60 + int(s)</pre>
In [24]:	
In [25]: Out[25]:	
	0 5912 40 1031602 8523 363 1095 2235 37 6 1 2741 2 1707 56 2 6 207 30 4 2 8138 1 2023 25 0 2 1905 16 3 3 9005 6 620860 777 161 153 1952 22 8
	4 122 1 666 1 0 0 1783 31 4 VISUALISATION
	Individual plots
In [26]:	<pre>plt.hist(data_train["category"]) plt.show() plt.plot(data_train["adview"]) plt.show()</pre>
	2000 1000 1000 1200 1000 12000 14000
In [27]:	Remove videos with adview greater than 2000000 as outlier data_train = data_train[data_train["adview"]<2000000]
In [28]:	<pre>import seaborn as sns</pre>
In [30]:	<pre>corr = data_train.corr() sns.heatmap(corr, mask=np.zeros_like(corr, dtype=np.bool), cmap=sns.diverging_palette(220, 10, as_cmap=True), square=True, ax=ax, annot=True) plt.show()</pre>
	<pre><ipython-input-30-c41adfee4e44>:3: DeprecationWarning: `np.bool` is a deprecated alias for the builtin `bool`. To silence this warning, use `bool` by itself. Doing this will not modify any behavior and is safe. If you specifically wanted the numpy scalar type, use `np.bool_` here. Deprecated in NumPy 1.20; for more details and guidance: https://numpy.org/devdocs/release/1.20.0-notes.html#deprecations sns.heatmap(corr, mask=np.zeros_like(corr, dtype=np.bool), cmap=sns.diverging_palette(220, 10, as_cmap=True), square=True, ax=ax, annot=True) vidid - 1</ipython-input-30-c41adfee4e44></pre>
	adview0.031
	likes0.13
	comment0.081
	duration - 0.0028 -0.012 0.0056 -0.035 0.00061 -0.033 -0.013 1 0.057 category0.039 -0.0049 0.033 -0.0092 0.011 -0.034 -0.026 0.057 1
	vidid adview views likes dislikes comment published duration category Split data
In [31]:	<pre>Y_train = pd.DataFrame(data_train.iloc[:, 1].values, columns = ['target']) data_train = data_train.drop(["adview"],axis=1) data_train = data_train.drop(["vidid"],axis=1) data_train.head()</pre>
Out[31]:	0 1031602 8523 363 1095 2235 37 6 1 1707 56 2 6 207 30 4
	2 2023 25 0 2 1905 16 3 3 620860 777 161 153 1952 22 8 4 666 1 0 0 1783 31 4
In [32]:	
Tn [22].	X_train, X_test, y_train, y_test = train_test_split(data_train, Y_train, test_size=0.2, random_state = 42)
<pre>In [33]: Out[33]:</pre>	<pre>X_train, X_test, y_train, y_test = train_test_split(data_train, Y_train, test_size=0.2, random_state = 42)</pre> <pre>X_train.shape</pre> (11708, 7)
	X_train, X_test, y_train, y_test = train_test_split(data_train, Y_train, test_size=0.2, random_state = 42) X_train.shape (11708, 7) Normalise Data from sklearn.preprocessing import MinMaxScaler scaler = MinMaxScaler() X_train = scaler.fit_transform(X_train)
Out[33]:	X_train, X_test, y_train, y_test = train_test_split(data_train, Y_train, test_size=0.2, random_state = 42) X_train.shape (11708, 7) Normalise Data from sklearn.preprocessing import MinMaxScaler scaler = MinMaxScaler()
Out[33]:	X_train, X_test, y_train, y_test = train_test_split(data_train, Y_train, test_size=0.2, random_state = 42) X_train.shape (11708, 7) Normalise Data from sklearn.preprocessing import MinMaxScaler scaler = MinMaxScaler() X_train = scaler.fit_transform(X_train) X_test = scaler.fit_transform(X_test) Evaluation Metrics
Out[33]: In [34]:	X_train, X_test, y_train, y_test = train_test_split(data_train, Y_train, test_size=0.2, random_state = 42) X_train.shape (11708, 7) Normalise Data from sklearn.preprocessing import MinMaxScaler scaler = MinMaxScaler() X_train = scaler.fit_transform(X_train) X_test = scaler.fit_transform(X_test) Evaluation Metrics from sklearn import metrics def print_error(X_test, y_test, model_name): prediction = model_name.predict(X_test) print('Mean Absolute Errori', metrics.mean_absolute_error(y_test, prediction))
Out[33]: In [34]:	X_train, X_test, y_train, y_test = train_test_split(data_train, Y_train, test_size=0.2, random_state = 42) X_train.shape (11788, 7) Normalise Data from sklearn.preprocessing import MinMaxScaler scaler = MinMaxScaler() X_train = scaler.fit_transform(X_train) X_test = scaler.fit_transform(X_test) Evaluation Metrics from sklearn import metrics def print_error(X_test, y_test, model_name): prediction = model_name.predict(X_test) print('Mean Assoulue Errore', metrics.mean_sbuslute_error(y_test, prediction)) print('Mean Squared Errore': netrics.mean_squared_error(y_test, prediction))) Linear Regression
Out[33]: In [34]:	X_train, X_test, y_train, y_test = train_test_split(data_train, Y_train, test_size=0.2, rendom_state = 42) X_train.shape (11788, 7) Normalise Data from sklearn_oregrocessing import MinMaxScaler scaler = MinMaxScaler() x_train = scaler.fit_transform(X_train) X_train = scaler.fit_transform(X_train) X_test = scaler.fit_transform(X_test) Evaluation Metrics from sklearn import metrics def print = correct(x_test, y_test, model_name): print('Mean Assolute Tenore', metrics.mean absolute_error(y_test, prediction)) print('Mean Assolute Tenore', metrics.mean absolute_error(y_test, prediction))) Linear Regression from sklearn import linear_model linear_regression() linear_regression = fit(X_train y_train) print('Mean fit in a model in a mode
Out[33]: In [34]:	X_train, X_test, y_train, y_test = train_test_split(data_train, Y_train, test_size=0.2, random_state = 42) X_train.shape (11788, 7) Normalise Data from sklearn.preprocessing import MinNavScaler scaler = MinNavScaler()
Out[33]: In [34]: In [50]:	X_train, X_test, y_train, y_test = train_test_split(data_train, Y_train, test_size=0.2, random_state = 42) X_train.shape (11708, 7) Normalise Data from sklearn.preprocessing import MinNaxScaler scaler = MinNaxScaler() X_train = scaler.fil_transform(X_train) X_trest = scaler.fil_transform(X_train) X_test = scaler.fil_transform(X_test) Evaluation Metrics from sklearn import metrics def print_error(X_test, y_test, model_name): prediction = model_name.predict(X_test) print(Man Squared Error); metrics.men_shoulte_error(y_test, prediction)) print(Man Squared Error); metrics.men_squared_error(y_test, prediction))) Linear Regression from sklearn import intoon.model inten_regression = intoon.model_linear.Regression() inten_regression.sit(X_test, y_test, innear_model_linear.prediction)) thean Absolute Error: 286.914249760125 thean Squared Error: 288.914249760125 thean Squared Error: 286.914249760125 thean Squared Error: 286.9144249760125 thean Squared Error: 286.914449760125 thean Squared Erro
Out[33]: In [34]: In [50]:	X_train, X_test, y_train, y_test = train_test_split(data_train, Y_train, test_size=0.2, random_state = 42) X_train.stape (11788, 7) Normalise Data from sileam.preprocessing_import Minkascaler scalar = RNUMScalar()
Out[33]: In [34]: In [50]:	S_Train, X_Train, y_Train, y_Train, y_Train, test_stands, v_Train, test_stands, nadds_state = 40)
Out[33]: In [34]: In [50]:	X_crain_X_crass_y_train_y_test = train_test_split(date_train_y_train_test_sizen0.2_nandom_state = 42) X_crain_stape (11787) Normalise Data
Out[33]: In [34]: In [50]:	X train, X tests, y train, y test = train test solitions train, Y train, test sized, 2, random state = 42) X train, shape (1278, 7) Normalise Data from silcorn, proprocessing import Mirroscolor states - Hirroscolor (1,778) A cost = coster (1,778) Train silcorn, import cost (1,778) A cost = coster (1,778) From silcorn (1,
Out[33]: In [34]: In [50]:	Commonwealth Comm
Out[33]: In [34]: In [50]:	Extends Extends Expension Expensio
Out[33]: In [34]: In [50]:	Extending Extending Extend