import imageio as iio import visvis as vv import numpy as np import matplotlib import matplotlib.pyplot as plt import pandas as pd from PIL import Image import csv import os %matplotlib inline pd.options.mode.chained assignment = None df = pd.read csv('savedwork.csv') In [9]: df country hash site\_nm date img\_url 2022-United f810500fcf4ffdbd89571a46 f810500fcf4ffdbd89571a469951c835 July https://assets.awwwards.com/awards/media/cache... 06-10 States Shaping Better 2022-147fd1ccb1a369fa1663d26dfff9bdd9 France https://assets.awwwards.com/awards/media/cache... 147fd1ccb1a369fa1663d20 Maritime 05-06 World No Fun 2022-United e077a8deb52eb3dc2560248fb6b2fb2f https://assets.awwwards.com/awards/media/cache... e077a8deb52eb3dc2560248 Studio 03-30 States Dynamic 2022-9313bd9422860e60229e52f2f5f8dafd https://assets.awwwards.com/awards/media/cache... 9313bd9422860e60229e52 Germany Solution 03-29 2022-Czech f14a4c67cadf1cfb5d6b9cb8ec2c432a https://assets.awwwards.com/awards/media/cache... f14a4c67cadf1cfb5d6b9cb8 **AllCaps** 03-28 Republic UNLOCK 2019-27971fd45b0cd48d79a7323c1a170203 Japan https://assets.awwwards.com/awards/media/cache... 27971fd45b0cd48d79a7323 YOU | WW 12-09 The Scott 2019-United 9811 ab5ea49896c0534621131fdaac491c15 Resort & ab5ea49896c0534621131fd https://assets.awwwards.com/awards/media/cache... 12-08 Kingdom Spa Sacha 2019bfb60a6e07853064a377f4c7d5f5a6d6 bfb60a6e07853064a377f4c Tourtoulou https://assets.awwwards.com/awards/media/cache... 12-07 - Portfolio The Geek 2019-United 799957330d0cc8fb36c3439cd9252688 https://assets.awwwards.com/awards/media/cache... 799957330d0cc8fb36c3439c States Designer 12-06 Limnia Modular 2019-United 9814 8fdef32d6f960f42baabcb3b6f331353 https://assets.awwwards.com/awards/media/cache... 8fdef32d6f960f42baabcb3l 12-05 States Fine Jewelry 9815 rows × 11 columns Formatting Dates to Date Types print(df['date'].dtype) df['date'] = pd.to\_datetime(df['date']) print(df['date'].dtype) object datetime64[ns] **Creating Date Parts** df['weekday'] = df['date'].dt.day\_of\_week df['day'] = df['date'].dt.day df['year'] = df['date'].dt.year df['month'] = df['date'].dt.month Out[38]: 2021 4358 2020 4086 2022 1131 2019 240 Name: year, dtype: int64 **Reviewing Weekday Trends** weekday\_df = df.groupby(['weekday','winner'])['year'].count().unstack('winner').reset\_index().set\_index('weekday') weekday\_df.index.names = [None] weekday\_df.head() ax = weekday\_df[[0.0,1.0]].plot(kind='bar', stacked=True, title = 'Weekday by Winners Count') Weekday by Winners Count 2000 winner 0.0 1750 1.0 1500 1250 1000 750 500 250 0 Need to Remove Weekdays of 5 & 6 as these are have winner details but no nominees weekday df Out[86]: winner 0.0 1.0 **0** 1846 126 **1** 1784 128 **2** 1757 128 **3** 1752 130 **4** 1770 129 2 127 7 129 print("DF length before removing Saturdays: "+str(len(df))) df = df[(df['weekday']!=5)]print("DF length after removing Saturdays: "+str(len(df))) df = df[(df['weekday']!=6)]print("DF length after removing Sundays: "+str(len(df))) DF length before removing Saturdays: 9815 DF length after removing Saturdays: 9686 DF length after removing Sundays: 9550 weekday df = df.groupby(['weekday','winner'])['year'].count().unstack('winner').reset index().set index('weekda') weekday df.index.names = [None] weekday df.head() ax = weekday df[[0.0,1.0]].plot(kind='bar', stacked=True, title = 'Weekday by Winners Count') Weekday by Winners Count 2000 1750 1500 1250 1000 750 500 winner 0.0 250 1.0 year df = df.groupby(['year','winner'])['year'].count().unstack('winner').reset index().set index('year') year df.index.names = [None] ax = year df[[0.0,1.0]].plot(kind='bar', stacked=True, title = 'Year by Winners Count') Year by Winners Count winner 4000 0.0 1.0 3500 3000 2500 2000 1500 1000 500 2020 2021 In [94]: month\_df = df.groupby(['month','winner'])['month'].count().unstack('winner').reset index().set index('month') month df.index.names = [None] month df.head() ax = month\_df[[0.0,1.0]].plot(kind='bar', stacked=True, title = 'Month by Winners Count') Month by Winners Count winner 1000 0.0 1.0 800 600 400 200 day df = df.groupby(['day','winner'])['day'].count().unstack('winner').reset index().set index('day') day df.index.names = [None] day df.head() day\_df[[0.0,1.0]].plot(kind='bar', stacked=True, title = 'Day by Winners Count', figsize=(10, 5)) <AxesSubplot:title={'center':'Day by Winners Count'}> Day by Winners Count winner 350 0.0 1.0 250 200 150 100 50 Solving for Imbalanced Classes from matplotlib import pyplot from matplotlib.pyplot import figure print('Non-Winners Count: '+str(len(df[df['winner']==0]))) print('Winners Count:'+ str(len(df[df['winner']==1]))) df win = df[df['winner']==1] df nom = df[df['winner']==0] diff = len(df[df['winner']==0]) - len(df[df['winner']==1]) print("Diff: "+str(diff)) figure(figsize=(5, 5), dpi=80) df['winner'].value\_counts().plot(kind='bar') pyplot.xticks(rotation = 90) plt.title("Distribution of Winners before resampling") pyplot.show() Non-Winners Count: 8909 Winners Count:641 Diff: 8268 Distribution of Winners before resampling 8000 6000 4000 2000 0 Random Sampling with Replacement from random import choices df smpld = df win.iloc[np.random.randint(0, len(df win), size=diff)] print(len(df smpld)) df = pd.concat([df, df\_smpld], axis=0) 8268 print('Non-Winners Count: '+str(len(df[df['winner']==0]))) print('Winners Count:'+ str(len(df[df['winner']==1]))) diff = len(df[df['winner']==0]) - len(df[df['winner']==1]) print("Diff: "+str(diff)) figure(figsize=(5, 5), dpi=80) df['winner'].value counts().plot(kind='bar') pyplot.xticks(rotation = 90) plt.title("Distribution after resampling") pyplot.show() Non-Winners Count: 8909 Winners Count:8909 Diff: 0 Distribution after resampling 8000 6000 4000 2000 0 0.0 1.0 **Encode non-numeric features** from sklearn import preprocessing le = preprocessing.LabelEncoder() le.fit(df["country"]) df['country\_enc']=(le.transform(df["country"])) le.fit(df["average\_color"]) df['average\_color\_enc']=(le.transform(df["average\_color"])) le.fit(df["dominant color"]) df['dominant\_color\_enc']=(le.transform(df["dominant\_color"])) df.head() site\_nm hash date country img\_url img\_ United 2022f810500fcf4ffdbd89571a469951c835 https://assets.awwwards.com/awards/media/cache... f810500fcf4ffdbd89571a469951cl 06-10 States Shaping Better 2022-France https://assets.awwwards.com/awards/media/cache... 147fd1ccb1a369fa1663d26dfff9b 147fd1ccb1a369fa1663d26dfff9bdd9 Maritime 05-06 World No Fun 2022-United e077a8deb52eb3dc2560248fb6b2fb2f https://assets.awwwards.com/awards/media/cache... e077a8deb52eb3dc2560248fb6b2f Studio 03-30 States Dynamic 2022-9313bd9422860e60229e52f2f5f8dafd https://assets.awwwards.com/awards/media/cache... Germany 9313bd9422860e60229e52f2f5f8c Solution 03-29 2022f14a4c67cadf1cfb5d6b9cb8ec2c432a https://assets.awwwards.com/awards/media/cache...f14a4c67cadf1cfb5d6b9cb8ec2c4 AllCaps 03-28 Republic Remove unnecessary columns corr\_cols = ['winner', 'day','month','year','country\_enc','average\_color\_enc','dominant\_color\_enc', 'unique\_col cols = ['day','month','year','country\_enc','average\_color\_enc','dominant\_color\_enc', 'unique\_color\_count\_50'] data = df[cols]target = df['winner'] In [407... data.shape (17818, 7) Out[407... In [408.. target.shape (17818,)Out[408... In [409... import seaborn as sn corrMatrix = df[corr\_cols].corr() sn.heatmap(corrMatrix, annot=True) plt.show() -0.0068 -0.018 | 0.049 | -0.043 | -0.035 | 1.7e-05 -0.041 winner - 0.8 0.017 -0.058 -0.013 -0.0037-0.0098 -0.012 - 0.6 -0.35 -0.0031 -0.022 -0.019 0.032 month -0.018 0.017 0.049 -0.058 -0.35 1 0.0025 -0.0053 0.022 0.046 - 0.4 -0.017 -0.037 -0.001 -0.043 -0.013 -0.0031 0.0025 country\_enc - 0.2 0.035 -0.0037 -0.022 -0.0053 -0.017 average\_color\_enc - 0.0 1.7e-05-0.0098 -0.019 0.022 -0.037 1 0.0014 dominant\_color\_enc unique\_color\_count\_50 -0.041 -0.012 0.032 0.046 -0.001 0.021 0.0014 average\_color\_enc dominant\_color\_enc In [410... from sklearn.model\_selection import train\_test\_split X\_train, X\_test, y\_train, y\_test = train\_test\_split( data, target, random\_state=0) In [411... print("X\_train shape:", X\_train.shape) print("y\_train shape:", y\_train.shape) X train shape: (13363, 7) y\_train shape: (13363,) In [412... print("X\_test shape:", X\_test.shape) print("y\_test shape:", y\_test.shape) X\_test shape: (4455, 7) y\_test shape: (4455,) In [413... from sklearn.neighbors import KNeighborsClassifier knn = KNeighborsClassifier(n\_neighbors=3) knn.fit(X\_train, y\_train) Out[413... KNeighborsClassifier(n\_neighbors=3) In [414... print("Model score:") print(knn.score(X\_test,y\_test)) Model score: 0.9099887766554433 In [415... no\_neighbors = np.arange(2, 20) train\_accuracy = np.empty(len(no\_neighbors)) test\_accuracy = np.empty(len(no\_neighbors)) In [416.. data.columns Out[416... Index(['day', 'month', 'year', 'country\_enc', 'average\_color\_enc', 'dominant\_color\_enc', 'unique\_color\_count\_50'], dtype='object') In [424... for i, k in enumerate(no\_neighbors): knn = KNeighborsClassifier(n\_neighbors=k) knn.fit(X\_train,y\_train) # Compute accuracy on the training set train\_accuracy[i] = knn.score(X\_train, y\_train) # Compute accuracy on the testing set test\_accuracy[i] = knn.score(X\_test, y\_test) # Visualization of k values vs accuracy plt.title('k-NN: Varying Number of Neighbors') plt.plot(no\_neighbors, test\_accuracy, label = 'Testing Accuracy') plt.plot(no\_neighbors, train\_accuracy, label = 'Training Accuracy') plt.legend() plt.xlabel('Number of Neighbors') plt.ylabel('Accuracy') plt.show() n=0 for i in test\_accuracy: print(str(no\_neighbors[n])+": "+str(round(i,4))) n = n+1k-NN: Varying Number of Neighbors 1.00 Testing Accuracy Training Accuracy 0.95 0.90 0.85 0.80 0.75 0.70 Number of Neighbors 2: 0.9549 3: 0.91 4: 0.91 5: 0.8716 6: 0.8691 7: 0.8368 8: 0.8352 9: 0.8067 10: 0.7969 11: 0.7708 12: 0.7603 13: 0.7407 14: 0.7311 15: 0.7149 16: 0.7051 17: 0.6947 18: 0.6893 19: 0.686 In [432... from sklearn.metrics import classification\_report  $\textbf{from} \ \, \texttt{sklearn.metrics} \ \, \textbf{import} \ \, \texttt{confusion\_matrix}$ model = KNeighborsClassifier(n\_neighbors = 2) model.fit(X\_train, y\_train) predictions = model.predict(X\_test) print(classification\_report(y\_test, predictions)) precision recall f1-score support 0.0 1.00 0.91 0.95 2204 1.00 0.92 1.0 0.96 2251 accuracy 0.95 4455 macro avg 0.96 0.95 0.95 4455 0.95 weighted avg 0.96 0.95 4455