In [267]: import numpy as np import pandas as pd import seaborn as sns from matplotlib import pyplot as plt %matplotlib inline Load the data In [268]: traindf = pd.read csv("train.csv", encoding="utf-8") traindf.head() # sibsp: siblings / spouses aboard the Titanic # parch: parents / children aboard the Titanic Out[268]: Passengerld Survived Pclass Name Sex Age SibSp Parch Ticket Fare Ca Braund. Mr. Owen male 22.0 7.2500 21171 Harris Cumings, Mrs. John Bradley 1 2 1 female 38.0 1 0 PC 17599 71.2833 (Florence Briggs Th... Heikkinen, STON/O2. 7.9250 2 3 Miss. female 26.0 3101282 Laina Futrelle. Jacques 113803 53.1000 C1 3 female 35.0 1 0 Heath (Lily May Peel) Allen, Mr. 8.0500 William 373450 male 35.0 Henry testdf = pd.read csv("test.csv", encoding="utf-8") In [269]: testdf.head() Out[269]: Passengerld Pclass Sex Age SibSp Parch Fare Cabin Embark Name **Ticket** Kelly, Mr. 0 892 male 34.5 0 330911 7.8292 NaN James Wilkes, Mrs. 893 James female 47.0 1 363272 7.0000 NaN (Ellen Needs) Myles, Mr. 2 894 Thomas male 62.0 240276 9.6875 NaN Francis Wirz. Mr. 895 3 male 27.0 315154 8.6625 NaN Albert Hirvonen. Mrs. 896 3 Alexander female 22.0 1 3101298 12.2875 (Helga E Lindqvist) In [270]: print(traindf.shape) print(testdf.shape) (891, 12)(418, 11) **Exploratory data analysis** Relationship between survived & Sex In [271]: sns.countplot(traindf['Sex'], hue = traindf['Survived']) Out[271]: <matplotlib.axes. subplots.AxesSubplot at 0x1a265bd588> Survived 0 400 1 300 200 100 male female Sex Relationship between survived & Embarked, Pclass, Sex In [272]: FacetGrid = sns.FacetGrid(traindf, row='Embarked', size=4.5, aspect=1.6) FacetGrid.map(sns.pointplot, 'Pclass', 'Survived', 'Sex', palette=None, order =None, hue order=None) FacetGrid.add legend() /anaconda3/lib/python3.6/site-packages/scipy/stats/stats.py:1706: FutureWar ning: Using a non-tuple sequence for multidimensional indexing is deprecate d; use `arr[tuple(seq)]` instead of `arr[seq]`. In the future this will be interpreted as an array index, `arr[np.array(seq)]`, which will result eith er in an error or a different result. return np.add.reduce(sorted[indexer] * weights, axis=axis) / sumval Out[272]: <seaborn.axisgrid.FacetGrid at 0x1a265bf080> Embarked = S1.0 0.8 0.6 0.4 0.2 0.0 Embarked = C1.0 0.8 0.6 Survived female male 0.4 0.2 0.0 Embarked = Q1.0 0.8 0.6 0.4 0.2 0.0 Pclass Relatinoship between Pclass and Survived In [273]: sns.barplot(x='Pclass', y='Survived', data=traindf) /anaconda3/lib/python3.6/site-packages/scipy/stats/stats.py:1706: FutureWar ning: Using a non-tuple sequence for multidimensional indexing is deprecate d; use `arr[tuple(seq)]` instead of `arr[seq]`. In the future this will be interpreted as an array index, `arr[np.array(seq)]`, which will result eith er in an error or a different result. return np.add.reduce(sorted[indexer] * weights, axis=axis) / sumval Out[273]: <matplotlib.axes. subplots.AxesSubplot at 0x1a2682fef0> 0.7 0.6 0.5 0.4 0.3 0.2 0.1 0.0 Pclass Relatinoship between Pclass, Age and Survived In [274]: grid = sns.FacetGrid(traindf, col='Survived', row='Pclass', size=2.2, aspect= grid.map(plt.hist, 'Age', alpha=.5, bins=20) grid.add legend() Out[274]: <seaborn.axisgrid.FacetGrid at 0x1a267d8710> Pclass = 1 | Survived = 0 Pclass = 1 | Survived = 1 40 30 20 10 Pclass = 2 | Survived = 0 Pclass = 2 | Survived = 1 40 30 20 10 0 Pclass = 3 | Survived = 0 Pclass = 3 | Survived = 1 40 30 20 10 0 20 40 60 80 40 80 **Handle Missing Values** Concat traindf & testdf In [275]: df = traindf.append(testdf) df = df.drop('Survived',axis=1) print(df.shape) (1309, 11)In [276]: df.info() <class 'pandas.core.frame.DataFrame'> Int64Index: 1309 entries, 0 to 417 Data columns (total 11 columns): Age 1046 non-null float64 295 non-null object Cabin Embarked 1307 non-null object Fare 1308 non-null float64 1309 non-null object Name Parch 1309 non-null int64 PassengerId 1309 non-null int64 Pclass 1309 non-null int64 Sex 1309 non-null object 1309 non-null int64 SibSp Ticket 1309 non-null object dtypes: float64(2), int64(4), object(5) memory usage: 122.7+ KB In [277]: | df.describe(include = 'all') Cabin Embarked **Fare** Name Parch Passengerld count 1046.000000 295 1307 1308.000000 1309 1309.000000 1309.000000 1309.0 unique 186 3 NaN 1307 NaN NaN NaN C23 Connolly, C25 S NaN NaN Miss. NaN NaN top C27 Kate NaN 6 914 NaN 2 NaN NaN freq 29.881138 33.295479 0.385027 655.000000 2. NaN NaN NaN mean 14.413493 NaN 51.758668 NaN 0.865560 378.020061 0.8 std NaN 0.000000 1.000000 min 0.170000 NaN NaN 0.000000 NaN 1. 328.000000 25% 21.000000 NaN NaN 7.895800 NaN 0.000000 2. 50% 28.000000 NaN NaN 14.454200 NaN 0.000000 655.000000 3. 39.000000 0.000000 75% NaN 31.275000 982.000000 NaN NaN 3. max 80.000000 NaN 512.329200 9.000000 1309.000000 3. In [278]: df.isna().sum() Out[278]: Age 263 1014 Cabin Embarked Fare Name Parch PassengerId Pclass Sex SibSp 0 Ticket 0 dtype: int64 In [279]: total = df.isna().sum().sort_values(ascending=False) percent 1 = df.isna().sum()/df.isna().count()*100 percent_2 = (round(percent_1, 2)) In [280]: missing data = pd.concat([total, percent 2], axis=1, keys=['Total', '%']) Out[280]: Total % 263 20.09 Age Cabin 1014 77.46 **Embarked** 0.15 Fare 1 0.08 0.00 Name Parch 0.00 **PassengerId** 0.00 **Pclass** 0.00 0.00 Sex 0.00 SibSp **Ticket** 0.00 Fill missing value- numerical In [281]: # fill numericla missing values with median In [282]: med = df.median() df = df.fillna(med) # Only apply to columns belong to numerical type df.isna().sum() Out[282]: Age 0 Cabin 1014 Embarked 2 0 Fare 0 Name Parch 0 PassengerId Pclass Sex SibSp Ticket dtype: int64 Fill missing value- categorical In [283]: # fill missing values (Embarked) with the most common one In [284]: | df["Embarked"].value counts() Out[284]: S 914 С 270 123 Name: Embarked, dtype: int64 In [285]: most = df["Embarked"].value counts().idxmax() most Out[285]: 'S' df["Embarked"] = df["Embarked"].fillna(most) In [286]: df.isna().sum() Out[286]: Age Cabin 1014 Embarked 0 0 Fare Name Parch PassengerId Pclass Sex SibSp Ticket dtype: int64 **One-Hot Encoding** In [287]: dummy = pd.get dummies(df["Embarked"]) df = pd.concat([df, dummy], axis=1) df.head() Out[287]: Τi Age Cabin Embarked Fare Name Parch Passengerld Pclass Sex SibSp Braund. Mr. Owen 0 22.0 NaN 7.2500 male 2 Harris Cumings, Mrs. John Bradley **1** 38.0 C 71.2833 1 female 1 PC 17 (Florence Th... Heikkinen, STON **2** 26.0 NaN 7.9250 Miss. 3 female 310 Laina Futrelle, Mrs. Jacques 1 female **3** 35.0 C123 S 53.1000 0 110 Heath (Lily May Peel) Allen, Mr. 373 **4** 35.0 NaN 8.0500 William 0 male 0 Henry In [288]: dummy = pd.get dummies(df["Sex"]) df = pd.concat([df, dummy], axis=1) df.head() Out[288]: Τi Cabin Embarked Fare Name Parch Passengerld Pclass Sex SibSp Braund, 22.0 NaN 7.2500 Mr. Owen 0 male 1 2 Harris Cumings, Mrs. John Bradley 1 female **1** 38.0 C85 C 71.2833 2 1 PC 17 (Florence Briggs Th... Heikkinen, STON 0 **2** 26.0 7.9250 3 NaN Miss. 3 female 310 Laina Futrelle, Mrs. Jacques **3** 35.0 C123 S 53.1000 110 0 1 female Heath (Lily May Peel) Allen, Mr. 8.0500 0 373 **4** 35.0 NaN William 5 0 3 male Henry In [289]: # Name -> title In [290]: def nameflow(s): return s.split(",")[-1].split(".")[0].replace(" ", "") df["title"] = df["Name"].apply(nameflow) df.head() Out[290]: Age Τi Cabin Embarked Fare Name Parch Passengerld Pclass Sex SibSp Braund, 0 22.0 NaN 7.2500 Mr. Owen male 1 2 Harris Cumings, Mrs. John Bradley **1** 38.0 C85 C 71.2833 2 1 PC 17 1 female (Florence Briggs Th... Heikkinen, STON **2** 26.0 NaN 7.9250 3 Miss. 3 female 310 Laina Futrelle, Mrs. Jacques **3** 35.0 C123 S 53.1000 110 1 female Heath (Lily May Peel) Allen, Mr. 373 **4** 35.0 8.0500 William 0 0 NaN male Henry In [291]: df["title"].value counts() Out[291]: Mr 757 260 Miss 197 Master 61 8 Dr Rev 8 Col 4 2 Ms Mlle 2 2 Major Capt 1 1 Lady theCountess 1 1 Jonkheer 1 1 Dona Sir 1 1 Don Name: title, dtype: int64 In [292]: # pd.crosstab(df["Survived"], traindf["Name"].apply(nameflow)) In [293]: def nameflow(s): s = s.split(",")[-1].split(".")[0].replace(" ", "") saved = ["Mr", "Mrs", "Miss"] if s in saved: return s else: return "X" s = df["Name"].apply(nameflow) s.head(10)Out[293]: 0 Mr1 2 Miss 3 Mrs 4 Mr5 6 Mr 7 X 8 Mrs 9 Mrs Name: Name, dtype: object In [294]: pd.set option('display.max columns', 50) In [295]: dummy = pd.get dummies(s) df = pd.concat([df, dummy], axis=1) df.head() Out[295]: Age Cabin Embarked Name Parch Passengerld Pclass Sex SibSp Τi Fare Braund, Mr. Owen **0** 22.0 NaN 7.2500 0 1 3 male 2 Harris Cumings, Mrs. John Bradley **1** 38.0 1 female C85 C 71.2833 1 PC 17 (Florence Briggs Th... Heikkinen, STON 3 female **2** 26.0 7.9250 Miss. 310 Laina Futrelle, Mrs. Jacques S 53.1000 **3** 35.0 C123 110 1 female Heath (Lily May Peel) Allen, Mr. 373 **4** 35.0 NaN 8.0500 0 William male Henry **Drop Columns** In [296]: df.columns Out[296]: Index(['Age', 'Cabin', 'Embarked', 'Fare', 'Name', 'Parch', 'PassengerId', 'Pclass', 'Sex', 'SibSp', 'Ticket', 'C', 'Q', 'S', 'female', 'male', 'title', 'Miss', 'Mr', 'Mrs', 'X'], dtype='object') In [297]: df = df.drop(['PassengerId', 'Name', 'Sex', 'Ticket', 'Cabin', 'Embarked', 'X' ,'title'], axis=1) df.head() Out[297]: Age Fare Parch Pclass SibSp C Q S female male Miss Mr Mrs 0 22.0 7.2500 1 0 0 1 0 **1** 38.0 71.2833 0 1 1 1 0 0 0 0 0 1 7.9250 0 0 0 1 **2** 26.0 0 **3** 35.0 53.1000 0 0 0 0 0 0 1 1 0 0 0 1 **4** 35.0 8.0500 3 0 0 1 0 create new feature In [298]: # Combine SibSp and Parch as a new feature called "relatives" #df['relatives'] = df['SibSp'] + df['Parch'] #df.loc[df['relatives'] > 0, 'not_alone'] = 0 #df.loc[df['relatives'] == 0, 'not_alone'] = 1 #df['not_alone'] = df['not_alone'].astype(int) In [299]: # To show people who have relatives # 1 : Have relatives # 0 : Not have relatives #df['not alone'].value counts() In [300]: df.head() Out[300]: Fare Parch Pclass SibSp C Q S female male Miss Mr Mrs Age 7.2500 0 22.0 0 0 1 0 **1** 38.0 71.2833 0 0 1 1 0 0 0 0 1 **2** 26.0 7.9250 0 0 0 1 0 **3** 35.0 53.1000 0 0 1 1 1 0 0 1 1 **4** 35.0 8.0500 0 0 0 1 Split df into train_df and test_df In [301]: df train = df[:len(traindf)] df test = df[len(traindf):] In [302]: df train['Survived'] = traindf['Survived'] df train.head() /anaconda3/lib/python3.6/site-packages/ipykernel launcher.py:1: SettingWith A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row_indexer,col_indexer] = value instead See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/ stable/indexing.html#indexing-view-versus-copy """Entry point for launching an IPython kernel. Out[302]: Fare Parch Pclass SibSp C Q S female male Miss Mr Mrs Survived Age **0** 22.0 7.2500 0 0 1 0 **1** 38.0 71.2833 0 0 1 1 0 0 0 0 1 **2** 26.0 7.9250 0 0 0 1 **3** 35.0 53.1000 0 1 1 0 0 1 0 0 0 1 1 **4** 35.0 8.0500 3 0 0 0 1 0 **Correlation of Features** plt.figure(figsize=(14, 11)) sns.heatmap(df_train.corr(), annot=True, cmap="RdBu") plt.title('Correlation of Features', y=1.05, size=15) Out[304]: Text(0.5,1.05,'Correlation of Features') Correlation of Features 0.097 -0.17 -0.23 0.03 -0.031 -0.0067 <mark>-0.081</mark> 0.081 -0.25 0.18 -0.065 - 0.8 Fare - 0.097 -0.12 -0.16 0.18 -0.18 0.12 -0.18 0.26 0.22 0.018 0.41 -0.011 -0.081 0.061 0.082 Parch - -0.17 0.25 -0.25 0.11 -0.33 0.23 0.018 0.083 -0.24 0.22 0.074 -0.13 0.13 -0.00058 0.14 -0.34 Pclass -0.34-0.15- 0.4 -0.06 -0.026 0.069 0.16 0.41 0.083 0.063 -0.035 SibSp · -0.23 0.11 -0.11 0.088 -0.25 C - 0.03 0.27 -0.011 -0.24 -0.06 -0.150.083 -0.083 0.026 -0.073 0.061 0.17 -0.074 0.17 -0.09 0.0037 Q - -0.031 -0.081 0.22 -0.026 -0.15 -0.078 S - -0.0067 -0.16 0.061 0.074 0.069 -0.12 0.12 -0.13 0.11 0.0027 -0.15 female - -0.081 0.18 0.25 -0.13 0.11 0.083 0.074 -0.12 male - 0.081 -0.18 -0.25 0.13 -0.11 -0.083 -0.074 0.12 -0.4 0.11 -0.00058 0.088 0.17 -0.2 0.33 Miss · -0.25 0.12 0.026 -0.130.18 0.11 -0.87 -0.18-0.33 0.14 -0.25 -0.073 -0.078 Mrs - 0.17 0.11 0.23 -0.15 0.063 0.061 -0.09 0.0027 -0.2 0.34 -0.8 Fare Parch Pclass SibSp Create a Decision tree to observe important features In [305]: from sklearn.model selection import train test split x = df train.drop(["Survived"], axis=1) y = df_train["Survived"] xtrain, xtest, ytrain, ytest = train_test_split(x, y, test_size=0.2) In [306]: **from sklearn.tree import** DecisionTreeClassifier clf = DecisionTreeClassifier(max_depth = 4, min_samples_leaf = 20) clf = clf.fit(xtrain, ytrain) In [307]: from sklearn.tree import export_graphviz import graphviz In [308]: dot_data = export_graphviz(clf, out_file=None, feature names=x.columns, class_names=["No", "Yes"], filled=True, rounded=Tru e, special characters=True) graph = graphviz.Source(dot_data) graph Out[308]: **Machine Learning model Random Forest** In [309]: from sklearn.ensemble import RandomForestClassifier from sklearn.metrics import accuracy_score In [310]: from sklearn.model_selection import GridSearchCV $params = {$ "n_estimators":range(20, 31), "max_depth":range(5, 11) clf = RandomForestClassifier() g = GridSearchCV(clf, params, cv=10) g.fit(x, y)print(g.best_params_) print(g.best_score_) {'max_depth': 6, 'n_estimators': 26} 0.8383838383838383 In [312]: | clf = RandomForestClassifier(max_depth= 6, n_estimators= 26) clf.fit(x, y)Out[312]: RandomForestClassifier(bootstrap=True, class_weight=None, criterion='gini', max_depth=6, max_features='auto', max_leaf_nodes=None, min_impurity_decrease=0.0, min_impurity_split=None, min_samples_leaf=1, min_samples_split=2, min_weight_fraction_leaf=0.0, n_estimators=26, n_jobs=1, oob_score=False, random_state=None, verbose=0, warm_start=False) In [313]: df_test.head() Out[313]: Age Fare Parch Pclass SibSp C Q S female male Miss Mr Mrs **0** 34.5 7.8292 0 0 1 0 **1** 47.0 7.0000 0 3 1 0 0 1 1 0 0 0 1 **2** 62.0 9.6875 0 2 0 0 1 0 0 0 0 1 **3** 27.0 8.6625 0 3 0 0 0 1 1 0 1 0 **4** 22.0 12.2875 3 1 0 0 1 0 0 0 In [314]: pre = clf.predict(df_test) pre Out[314]: array([0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 1, 0, 1, 1, 0, 0, 0, 1, 0, 1, 1, 0, 0, 0, 1, 0 1, 0, 1, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 1, 1, 0, 0, 1, 1, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 1, 1, 1, 0, 0, 1, 1, 0, 0, 0, 1, 0, 0, 1, 0, 1, 1, 0, 0, 0, 0, 0, 1, 0, 1, 1, 1, 0, 1, 0, 0, 0, 1, 0, 1, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 1, 1, 1, 0, 0, 1, 0, 1, 1, 0, 1, 0, 0, 1, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 1, 1, 0, 1, 1, 1, 1, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 1, 0, 1, 1, 0, 0, 1, 0, 1, 0, 1, 0, 0, 0, 0, 0, 1, 0, 1, 0, 1, 1, 0, 1, 1, 1, 1, 1, 0, 0, 1, 0, 1, 0, 0, 0, 0, 1, 0, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 1, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 1, 1, 1, 0, 0, 0, 0, 1, 0, 1, 1, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 1, 0, 0, 0, 0, 1, 1, 1, 1, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 1, 0, 1, 0, 0, 0, 0, 0, 1, 1, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 1, 0, 1, 0, 1, 0, 1, 1, 0, 0, 0, 1, 0, 1, 0, 0, 1, 0, 1, 1, 0, 1, 0, 0, 1, 1, 0, 0, 1, 0, 0, 1, 1, 0, 0, 0, 0, 0, 1, 1, 0, 1, 0, 0, 0, 0, 1, 0, 0, 1, 0, 1, 0, 0, 1, 0, 1, 0, 0, 0, 0, 1, 1, 1, 1, 1, 0, 1, 0, 0, 1]) In [315]: resultdf = pd.DataFrame() resultdf["PassengerId"] = testdf["PassengerId"] resultdf["Survived"] = pre In [316]: resultdf.head() Out[316]: Passengerld Survived 0 892 0 1 893 0 2 894 0 3 895 0 896 resultdf.to_csv("result.csv", encoding="utf-8", index=False) In [317]:

In [318]:

1s