In [49]: from sklearn.ensemble import RandomForestClassifier from sklearn.datasets import make classification from sklearn.model selection import GridSearchCV import pandas as pd import numpy as np from sklearn.metrics import confusion matrix datas = pd.read csv('/Users/uliaandreeva/Desktop/heart.csv') In [24]: datas.head(10) age sex cp trestbps chol fbs restecg thalach exang oldpeak slope ca thal Out[24]: target 0 63 1 3 145 233 1 0 150 0 2.3 0 0 1 1 1 37 2 130 250 187 3.5 0 2 1 2 41 0 1 130 204 0 0 172 0 1.4 2 0 2 1 236 2 3 56 1 1 120 0 1 178 0 0.8 2 0 1 2 4 57 0 0 120 354 0 1 163 1 0.6 2 0 1 140 192 0 148 0 1 0.4 2 6 56 0 1 140 294 0 0 153 0 1.3 1 0 1 7 44 1 1 120 263 0 1 173 0 0.0 2 0 3 1 8 52 1 2 172 199 1 1 162 0 0.5 2 0 3 1 57 150 174 2 1 168 In [14]: np.random.seed(0) rf = RandomForestClassifier(10, max depth=5) y_train = datas['target'] X train = datas.drop(['target'], axis=1) params = { "n_estimators": [10, 20, 30], 'max_depth' : range(3, 8), 'min_samples_split': [2, 6, 10], 'min_samples_leaf' : [1, 5, 10],} search = GridSearchCV(rf, params, cv=5) search.fit(X train, y train) Out[18]: GridSearchCV(cv=5, estimator=RandomForestClassifier(max depth=5, n estimators=10), param_grid={'max_depth': range(3, 8), 'min samples_leaf': [1, 5, 10], 'min samples split': [2, 6, 10], 'n estimators': [10, 20, 30]}) rf = search.best estimator imp = pd.DataFrame(rf.feature_importances_, index=X_train.columns, columns=['importances_) print(imp.sort_values('importance', ascending=False).head(5)) importance 0.237248 са 0.185174 exang 0.145520 ср slope 0.094971 thal 0.071552 imp.sort_values('importance').plot(kind='barh', figsize=(12, 8)) Out[23]: <AxesSubplot:> ca exand ф slope thal thalach oldpeak trestbps age sex chol restecg importance 0.00 0.05 0.10 0.15 0.20 gribs = pd.read csv('/Users/uliaandreeva/Desktop/training mush.csv') gribs.head() stalkstalkcolorgillgillgillgillstalkcolorcapcapcapbruises odor surface color attachment spacing size color shape abovebelowring ring 0 2 0 1 0 0 9 7 3 5 1 1 3 1 2 0 4 0 0 10 4 5 1 2 2 3 0 2 0 7 4 3 0 0 3 0 5 0 2 0 7 2 5 10 6 4 3 3 1 1 0 0 1 3 5 rows × 23 columns rrf = RandomForestClassifier(random state=0) params = { "n_estimators": range(10, 51, 10), 'max_depth' : range(1, 13, 2), 'min_samples_split': range(2, 9, 2), 'min_samples_leaf' : range(1, 8),} x_train = gribs.drop(['class'], axis=1) y_train = gribs['class'] search = GridSearchCV(rrf, params, cv=3, n jobs=-1) search.fit(x_train, y_train) search.best_params_ Out[38]: {'max_depth': 9, 'min samples leaf': 1, 'min samples split': 2, 'n estimators': 10} rf = search.best_estimator_ In [42]: imp = pd.DataFrame(rf.feature_importances_, index=x_train.columns, columns=['importances_') print(imp.sort_values('importance', ascending=False).head(25)) importance odor 0.188376 gill-color 0.103861 stalk-root 0.103793 0.083564 spore-print-color gill-spacing 0.080840 0.070726 ring-type bruises 0.070109 gill-size 0.068461 0.048296 stalk-surface-below-ring population 0.043783 stalk-surface-above-ring 0.031802 habitat 0.022611 cap-color 0.021773 stalk-color-above-ring 0.017689 ring-number 0.013910 veil-color 0.010881 0.007563 stalk-shape stalk-color-below-ring 0.006253 0.003130 cap-surface cap-shape 0.002577 veil-type 0.000000 0.000000 gill-attachment In [45]: test data = pd.read csv('/Users/uliaandreeva/Desktop/testing mush.csv') test data.head() stalkstalk-Out[45]: capcapсарgillgill- gillgillstalksurfacecolorbruises odor shape surface color attachment spacing size color shape belowabovering ring 0 0 3 8 1 3 1 0 0 4 0 2 7 1 5 3 5 0 10 2 6 2 7 2 3 3 4 0 8 1 0 1 0 1 3 3 2 0 1 0 1 0 1 6 3 2 2 4 0 8 1 0 1 0 1 ... 1 6 5 rows × 22 columns In [47]: predictions = search.predict(test data) predictions Out[47]: array([0, 0, 1, ..., 1, 0, 1]) In [48]: answer = 0for i in predictions: **if** i == 1: answer += 1 else: answer = answer answer Out[48]: 976 right anw = pd.read csv('/Users/uliaandreeva/Desktop/testing y mush.csv') right anw.head(10) class 0 0 1 0 2 1 3 1 4 1 5 0 6 1 7 0 8 1 9 1 mmm = confusion matrix(right anw, predictions) mmm Out[53]: array([[1055, 0], 976]]) [0, import seaborn as sns In [54]: sns.heatmap(mmm, annot=True, annot kws={"size": 16}) Out[54]: <AxesSubplot:> - 1000 1.1e + 030 - 800 - 600 400 0 9.8e + 02200 0 1 ships = pd.read csv('/Users/uliaandreeva/Desktop/invasion.csv') ships.head() class g_reflection i_reflection speed brightness time_of_observance volume 0 transport 2.190672 6.716633 62.168208 0.347465 158221 44.932446 1 transport 3.453276 8.995909 62.994707 0.590094 385972 41.568300 446482 2 transport 2.432994 6.938691 62.245807 0.329288 40.123467 6.083763 3.019459 3 fighter 18.474555 0.174738 210125 11.384865 4 fighter 12.876769 2.452950 195.805771 0.150446 23109 11.328806 In [61]: ML ships = RandomForestClassifier(random state=0) search = GridSearchCV(ML_ships, params, cv=3) x train = ships.drop(['class'], axis=1) search.fit(x_train, ships['class']) search.best params Out[61]: {'max_depth': 3, 'min samples leaf': 1, 'min_samples_split': 2, 'n estimators': 10} In [62]: test_ships = pd.read_csv('/Users/uliaandreeva/Desktop/operative_information.csv') test ships.head() g_reflection i_reflection speed brightness time_of_observance volume 3.916691 513.954279 0 7.516543 0.177247 105908 13.267224 6.967689 1 4.322988 63.752970 0.545922 39.833130 277855 2 4.595724 9.098297 62.233948 0.389201 160662 42.014556 3 2.689675 7.964869 62.475495 0.541081 162092 42.056829 0.174757 8.075576 5.169719 336.441261 466853 11.779813 rf = search.best estimator y_pred = rf.predict(test ships) y_pred Out[63]: array(['fighter', 'transport', 'transport', ..., 'transport', 'fighter', 'transport'], dtype=object) s = pd.Series(y_pred) In [64]: s.groupby(s).count() Out[64]: cruiser 230 675 fighter transport 595 dtype: int64 In [65]: imp = pd.DataFrame(rf.feature_importances_, index=x_train.columns, columns=['importances_, index=x_train.columns=['importances_, index=x_train.columns=['importances print(imp.sort_values('importance', ascending=False).head(25)) importance brightness 0.363803 volume 0.257661 speed 0.171441 i_reflection 0.124372 g reflection 0.082723 time of observance 0.000000 space = pd.read csv('/Users/uliaandreeva/Desktop/space can be a dangerous place.csv') In [67]: space.head() phi peradventure_index dustiness black_hole_is_near buggers_were_noticed nearby_syst r 169.1 138.0 22.3212 0.706285 0 -0.410512 1 11.1 148.0 1.4652 1 201.0 274.6 36.2472 0.756457 1 1 172.8 173.0 22.8096 0.035221 1 223.3 222.0 0 29.4756 0.197271 1 sns.heatmap(space.corr(), annot=True) Out[68]: <AxesSubplot:> - 1.0 r- 1 0.001 0.00330.00230.00180.000330.0058 0.001 0.00190.00290.00470.0008 0.003 0.001 1 - 0.8 0.001 0.00330.00230.00180.000330.0058 peradventure_index - 0.6 dustiness -0.00330.00190.0033 0.00470.00130.00340.0005 1 black_hole_is_near -0.00230.00290.00230.0047 0.00190.0027 0.053 1 - 0.4 buggers_were_noticed -0.00180.00470.00180.00130.0019 -0.003 0.34 0.2 0.3 dangerous -0.0058 0.003 0.00580.0005 0.053 0.34 0.3 1 0.0 듄 buggers were noticed nearby_system_has_planemo peradventure index dustiness dangerous df.applymap() df.apply() df.transform()