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First there was an idea:

Find intersting dataset Apply gathered knowledge Make something usefull

No 1

We found mushrooms.csv from **kaggle**

(https://www.kaggle.com/uciml/mushroom-classification?select=mushrooms.csv)

No missing values

No duplicates

23 columns and 8124 unique rows

Each row = mushroom

Eeach column = mushroom's parameter

Columns provide information such as:

shape and color of different parts of mushroom

where mushroom may be found

how big is the population

odor

bruices

and main column - "class" (edible or poisonous)

All information is categorical.

"Hmm... it is good data set for project" - we thought

And decided to:

Clear the data (we did not know that this is already clear)

Train some models to predict edability

Find strongest correlations between edability and other parameters

Find intersting correlations among other attributes

Write a programm for mushroom gatherers based on trained model

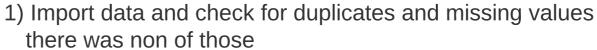


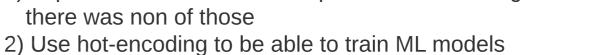
TOXIC

All tests led us to developing simple but usefull programm writen on python with use of pygame Programm asks questions about mushroom parameter and then predicts edability of mushroom using trained model



- Nº2





- map function and dummies turned our 23 columns into 114 columns 3) Use apriori algorithm to find interesting connections between attributes
- had not led to any interesting results
- 3) Run Lasso and Ridge regression with best alphas and watch, which attributes had high coefficients regression does not really help with classification, but model still search for useful corellations and try to give values as close as possible to 1 (edible) or 0 (poisonous)
- Both regression models was agree on 'odor' and 'spore color' parameters, but Lasso prefers 'stalk color' as 3rd most valuable parameter, Ridge prefered 'ring type'
- 4) Run KNN and RandomForest models, trained only on parameters with high coefficients from Lasso and Ridge regression

this led us to model with 99.7%(on 3 valuable parameters from Lasso) and 99.5% (on 3 valuable parameters from Ridge) accuracy.

Based on 'odor', 'spore-print-color' and 'stalk-color-below-ring' Knn prediction score: 0.9975381585425899

Random Forest prediction score: 0.9975381585425899

Based on 'odor', 'spore-print-color' and 'ring-type' Knn prediction score: 0.9950763170851797 Random Forest prediction score: 0.9950763170851797

which was pretty good, but not very usefull for mushroom gatherers. Odor is too subjective and spore color too hard to notice

- 5) Return to original data set and drop columns with parameters that are hard to define. such as 'odor', 'spore print color', 'population', etc. 13 columns left
- 6) Run Lasso and Ridge regression with parameters that are left after cleaning models were agree on 3 parameters, that are very easy to define, that are:





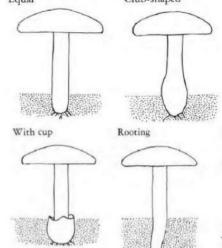


Stalk root shapes

We did not even try SVM

because encoded dataset

is too high dimensional



And prediction score reduces by less than 5%

Based on 'stalk-shape', 'gill-color' and 'stalk-root' Knn prediction score: 0.9533932951757972

Random Forest prediction score: 0.9533932951757972