GROWING (FOR) CONCERN

INTRODUCTION

Data Used:

SAVSNET Tumour Database

Tools Utilized:

- Python Pandas & Numpy for data manipulation
- Python Seaborn and Matplotlib for visualizations
- Python Scikit-learn for modeling and ML
- Python Seaborn and Matplotlib for visualizations



METHODOLOGY

- > Data Collection and Cleaning
 - Tumor data sourced from relatively new database of pet cancer cases provided by SAVSNET: Small Animal Veterinary Surveillance Network
- > Feature Engineering
 - > Segmented categorical data to try and determine indicators of multiple tumor count and potential high/low risk features
- Modeling and Selection
 - > Trained an ensemble voting classifier to determine likelihood of multiple lesions

PROCESS

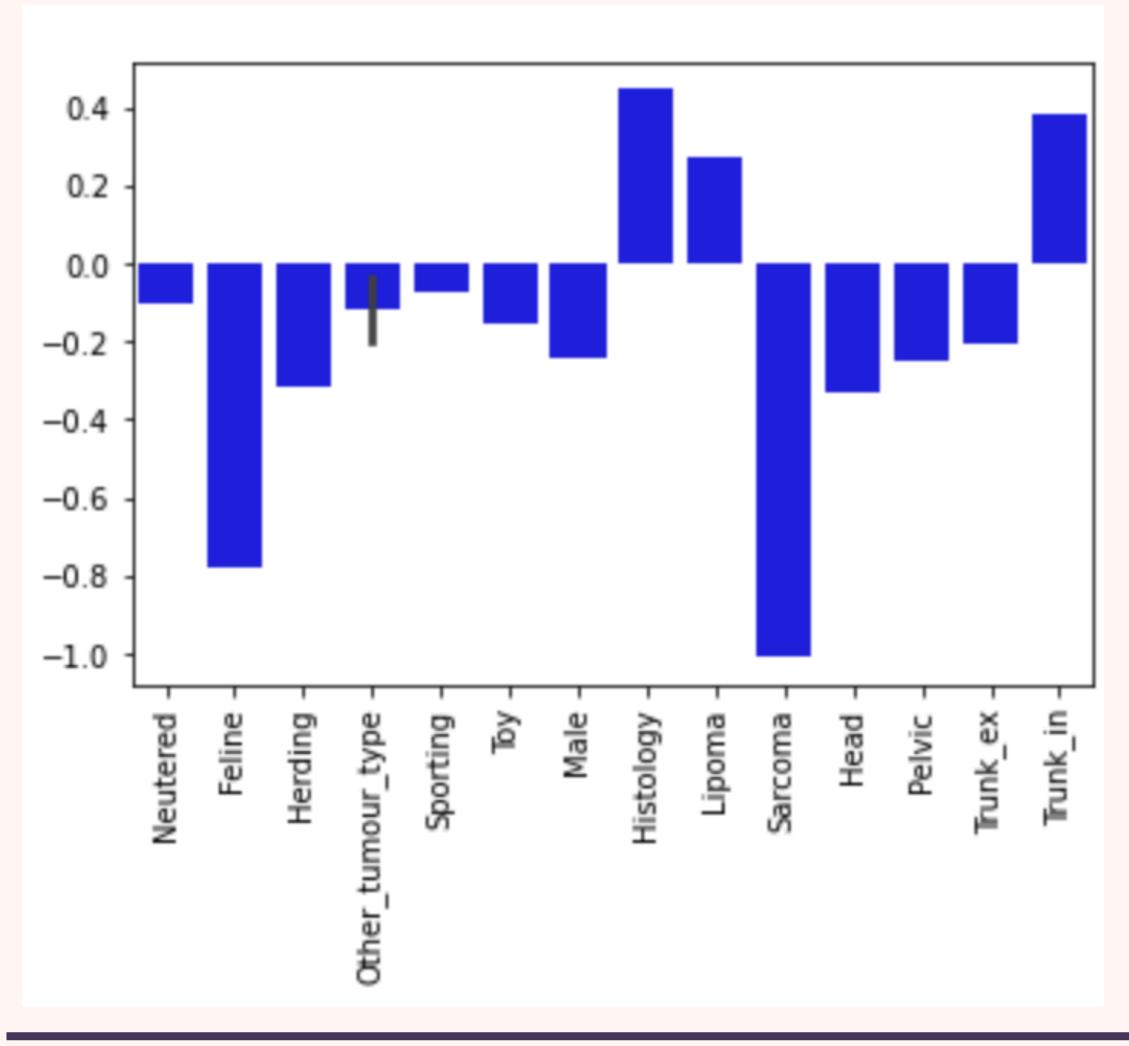
ORIGINAL DATA

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 109895 entries, 0 to 109894
Data columns (total 10 columns):
    Column
                            Non-Null Count
                                             Dtype
                            109895 non-null
                                             object
     Species
    Breed
                                             object
                            109895 non-null
                                             object
    Gender
                            109895 non-null
                            109895 non-null
                                             object
     Histo Cyto
                                             int64
     Tumours in the report
                            109895 non-null
    Primary_tumour
                            109895 non-null
                                             object
    Grade 2 tier
                                             object
                            13539 non-null
    Grade 3 tier
                                             object
                            9660 non-null
     Location
                                             object
                            108793 non-null
                            109895 non-null int64
     Neutered
```

Original dataset included various traits about prior cases including physical traits of the animal as well as how and where the tumor was found.

A vast majority of these needed further segmentation to become viable features, and a number were immediately removed (e.g. anonymous practice ID)

HIGHEST/LOWEST RISK FEATURES



- These features provided the final baseline for modeling and learning
- Notable features which did not provide a signal either way:
 - Non-sporting Breed Group
 - Lymphoma type Tumour
 - The Lymph System Location

RESULTS

DATA SELECTION & MODEL METRICS

<class 'pandas.core.frame.DataFrame'> RangeIndex: 109895 entries, 0 to 109894 Data columns (total 15 columns): # Column Non-Null Count Dtype Neutered 109895 non-null int64 Feline 109895 non-null uint8 Herding uint8 109895 non-null Other_tumour_type 109895 non-null uint8 Sporting 109895 non-null uint8 Toy 109895 non-null uint8 Male 109895 non-null uint8 Histology uint8 109895 non-null Lipoma 109895 non-null uint8 Other_tumour_type 109895 non-null uint8 109895 non-null 10 Sarcoma uint8 109895 non-null uint8 11 Head Pelvic 109895 non-null uint8 13 Trunk ex 109895 non-null uint8 Trunk in 109895 non-null uint8





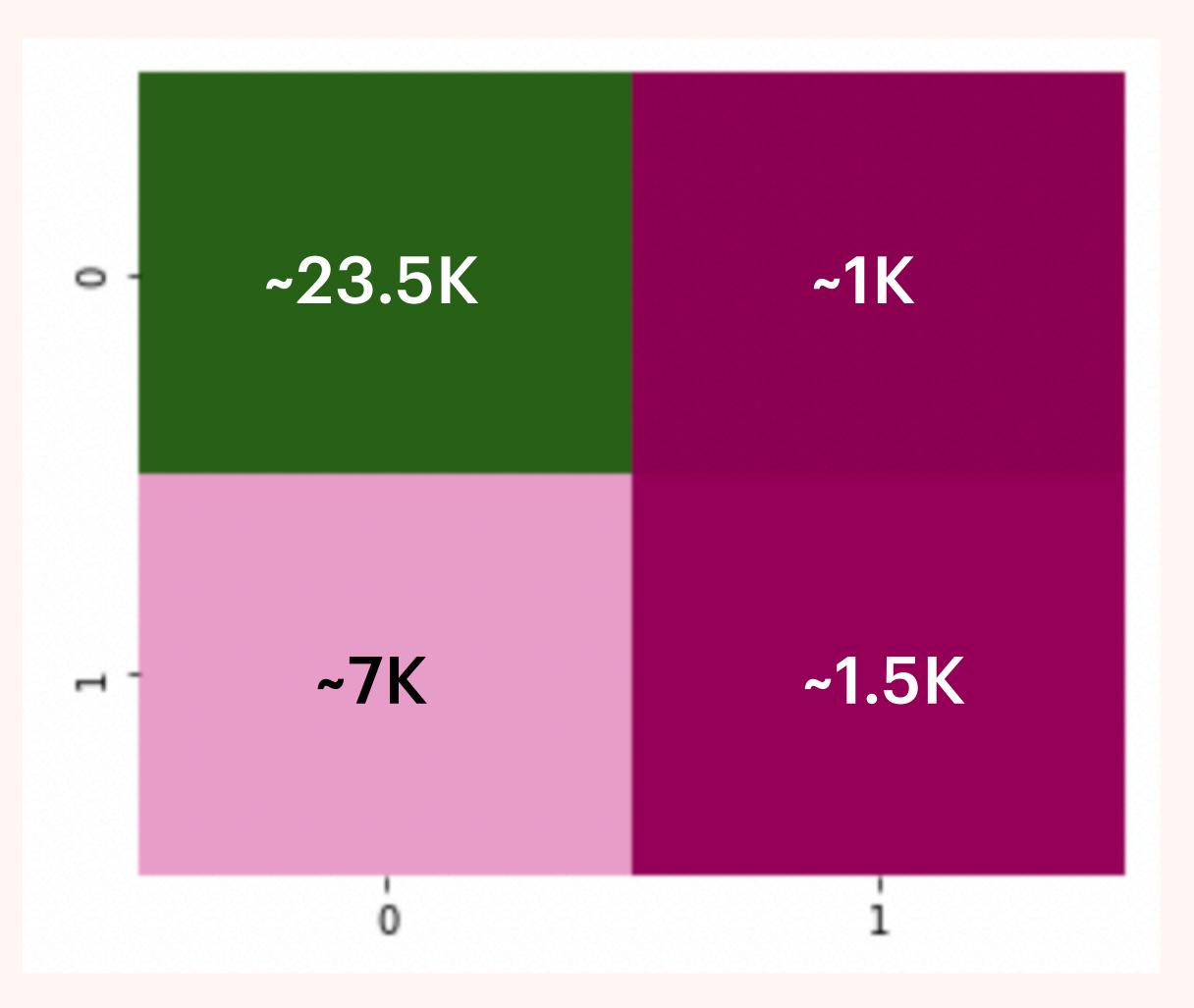


FINAL MODEL

Final Model

- Voting Classifier Ensemble
 - Decision tree weighted at 2 captures more true positives, but struggles to classify negatives accurately
 - Logistic Regression weighted at 1.4
 helps compensate for negative
 difficulties in Decision Tree

Accuracy of .7603 with highest confidence on both Positive and Negative predictions.



FURTHER ACTION AND NOTEABLES

Low Risk Features:

- > Felines significantly more likely to only have one lesion
- Sarcoma almost always only presents in one growth
- Herding Group dogs less likely to have multiple presentations

High Risk Features:

- Multi-growth Cancers are more likely to effect organs and connective tissue than external
- Lipoma is most likely to present in multiples
- Age not provided in data, but definitely would affect.

Further data acquisition concerning age could temper the results shown here, as increased age in all animals is linked to increase chance of cancer.

Further work with XGBoost would also likely assist in refining negative accuracy overall, and reduce false negatives