Shelter Animal Outcome Predicting

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Project Description

Project Goals

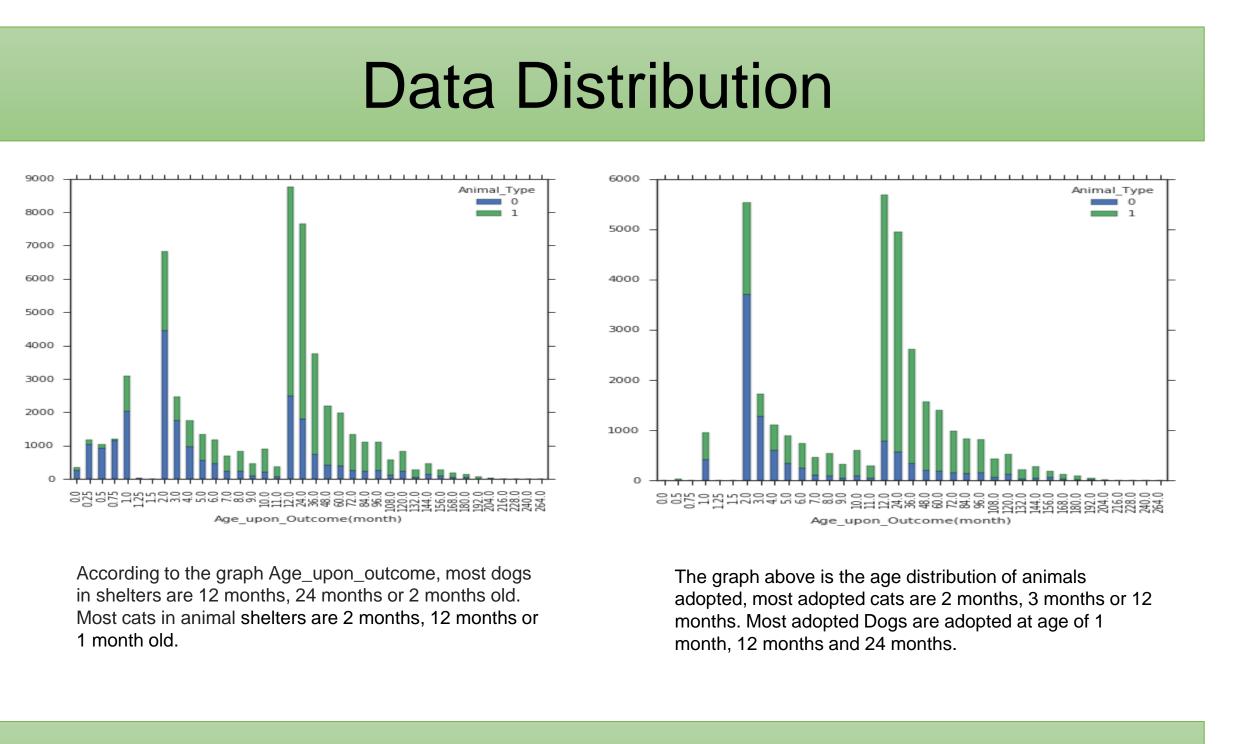
- The purpose of this project is to select the best model from various machine learning models such as decision tree, random forest,etc. to predict the outcome of shelter animals (Dog & Cat) based on information from Austin Animal Center Public Adoption Information.
- The dependent (target) variable for this project is the animal's outcome(Adopted, Transferring,Return to Owners, Died, Disposal or Euthanasia)
- The independent variables are animal types, sex, age, breed and color.

Data Exploration

- This dataset named "Shelter Animal Outcome Dataset" can be downloaded from Official City of Austin
- https://data.austintexas.gov/Health/Austin-Animal-Center-Outcomes/9t4d-g238
- This dataset has 54076 instances with three outcomes and 37016 instances with two outcomes and 5 independent variables which were extracted from Official City of Austin dataset.

Data Distribution

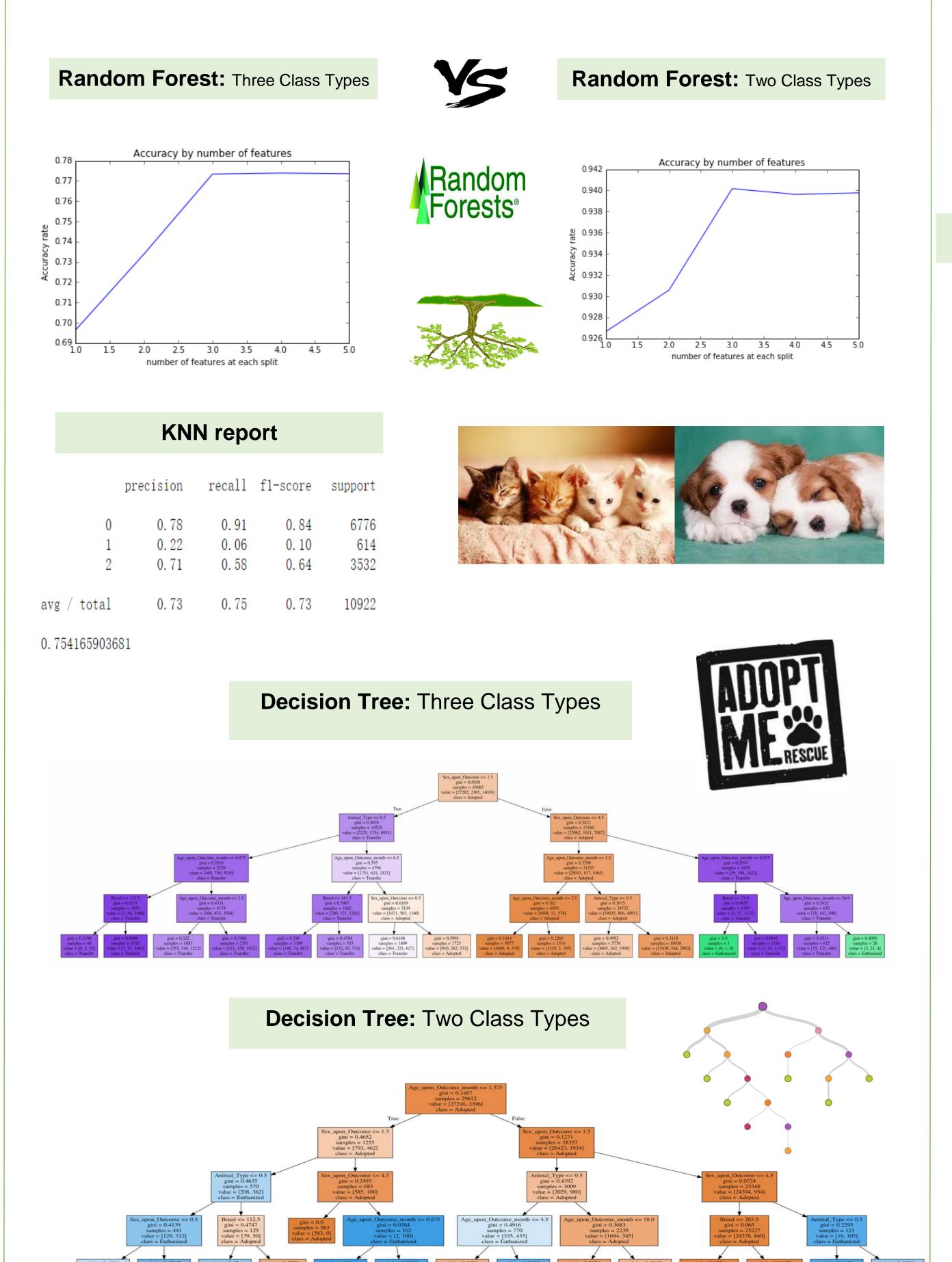




Machine Learning Medale



• Logistic regression measures the relationship between the control independent variables by estimating probabilities using a logistary reaction of the control independent variables by estimating probabilities using a logistary reaction of the control independent variables by estimating method for classification decision forests is an ensemble learning method for classification forests correct for decision trees' habit of own procession. Tree is a non-parametric supervised learning method create a model that predicts the value of a target variable by lefeatures. • k-Nearest Neighbor(KNN) is K nearest neighbors is a simple cases based on a similarity measure. • SVC is supervised learning models with associated learning all for classification and regression analysis (plot classification procession and regression analysis (plot classification procession for the control of the con	stic funcation to e of the verfitting of use learning algorithe robabilitize the D00, mu	nction, withat oper classed for classed fo	which is erates bes (classer train assificate decis at store at analyzer regres ass='mure recall for 10.94 0.01 0.58	s the cuby consistication ing set tion and ion rule as all avected	structing a on) of the on, of the	logistic distribution multitude of decis individual trees. ion. The goal is to d from the data
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		<u>6</u> 0.6)	
class 1		Precision				
		- 1	micro-a	verage Pr	ecision-recall	curve (area = 0.76)
	— Precision-recall curve of class 0 (area = 0.81)					
Probability						1 (area = 0.09) 2 (area = 0.68)
0.1 0.2 0.3 0.4 0.5 0.6 0.7 0.8 0.9		0.0 L 0.0			0.4 0.1	



Logistic Regression: Two Types Class

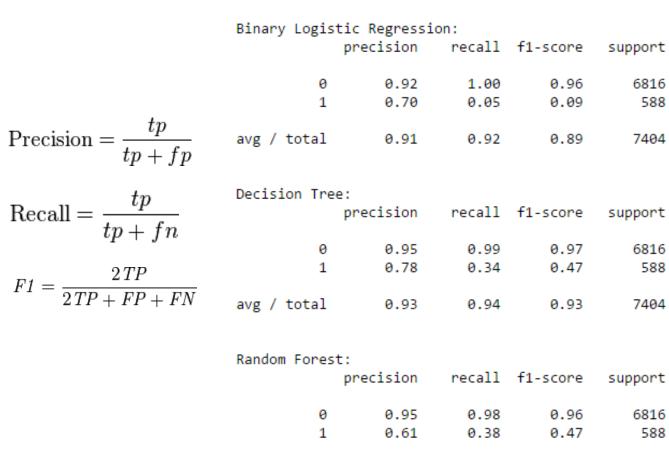
	Animal_Type	Sex_upon_Outcome	Age_upon_Outcome_month	Breed	Color
Coefficient	-1.1489275	-0.71967593	0.01079331	0.00135937	-0.00197791

Comparison Among Different Methods:



0.65 0.66 0.63 10922 precision recall f1-score support 0.01 0.02 0.57 0.65 3553

0.73 0.75 0.73 10922



Conclusions

- ❖ After comparison the result between different Machine Learning Method, it turns out that Animal Type and Sex_upon_Outcome have more influence than other variables on whether an animal will be adopted from shelter.
- ❖ In general, Dogs have a better chance to be adopted than cat.
- Neutered or spayed animals are more likely to be adopted, and if also take sex into account the likelihood of adoption is as follows: Spayed Female > Neutered Male> Intact Male> Intact
- ❖ Age, Breed and Color have a very small effect on adoption, which is a little contradicted to our common sense.





