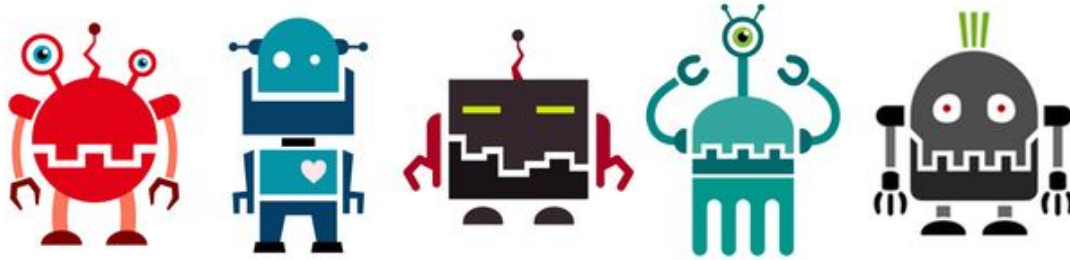
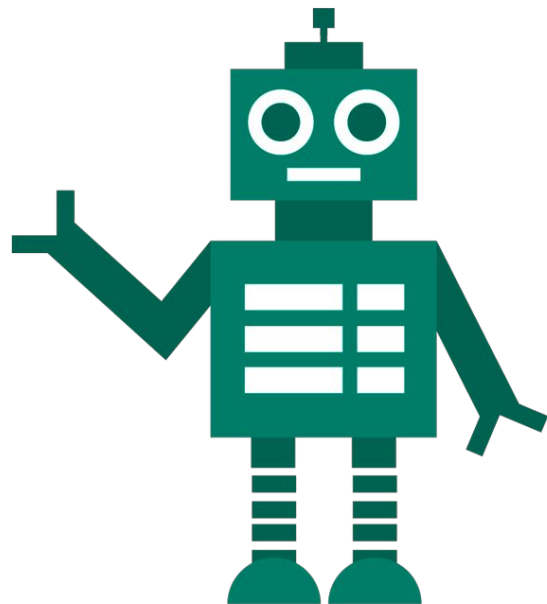


# Bots vs. Humans



# Action plan

- Project scope
- Data summary
- Models
- Findings
- Next steps



# Predict if an online bid is made by a machine or human



- 2015 Kaggle Competition
- 7.6 million bids on different auctions
- Use classification methods to distinguish between two classes

# Why bots are bad?



- Create a bad experience for real users
- Users less likely to come back to the auction site
- Over 95% of unique users are human

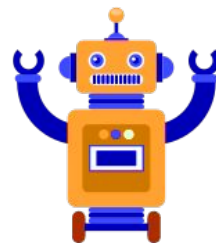
# Initial hiccups

## Class imbalance

- 7+ million bids
- 1910 real users
- 103 bots

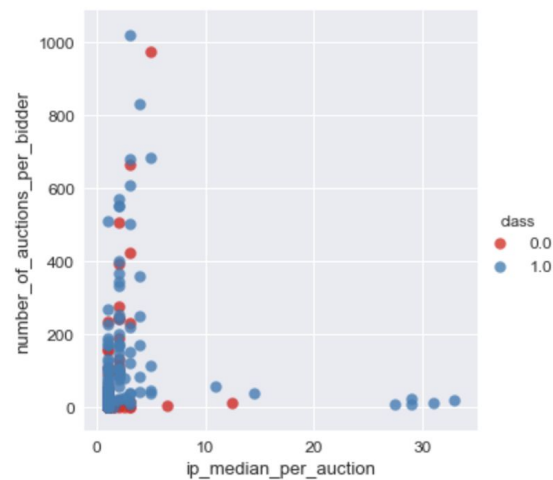
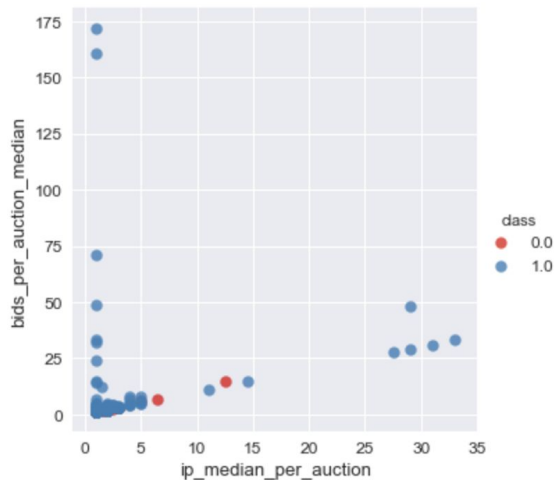
## Time obfuscation

- Time stamps have been hidden
- Ex. 9709222052631578
- Time is essential in spotting bots



# Modeled data

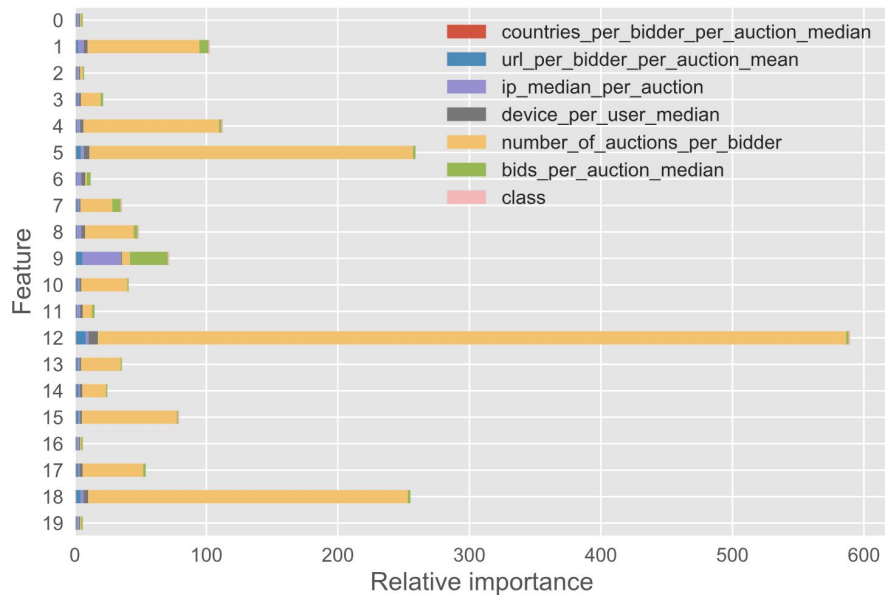
- SVM basic
- SVM RBF kernel
- KNN, GridSearchCV



# Model results

- RBF kernel gives best accuracy scores (0.878)
- Linear split accuracy is lower (.80)
- Best KNN result on K=5 ( $\sim$ .83)

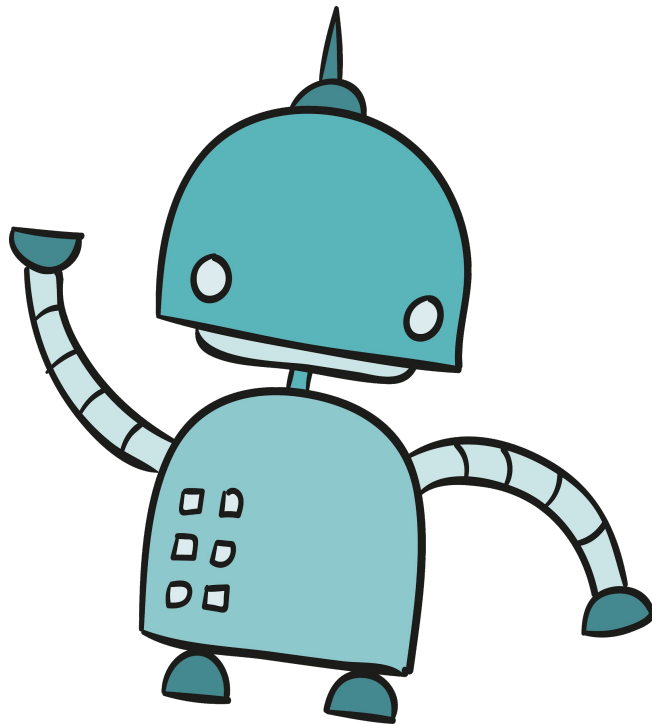
# Importance summary





# Further Steps

- Oversampling the Bots class
- Time stamp
- Other models
  - Random Forest
  - Various Kernels



# Thank you!

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**GitHub**



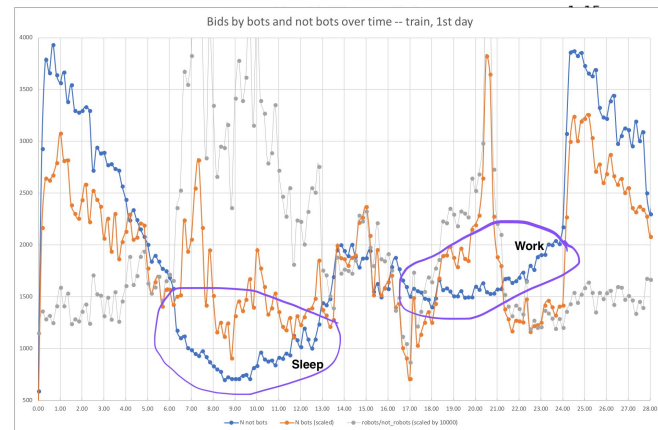
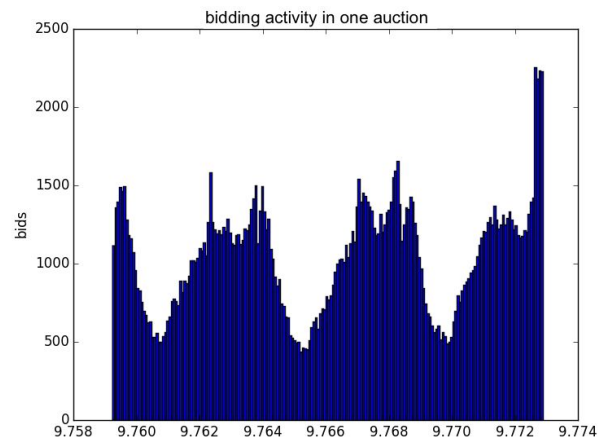
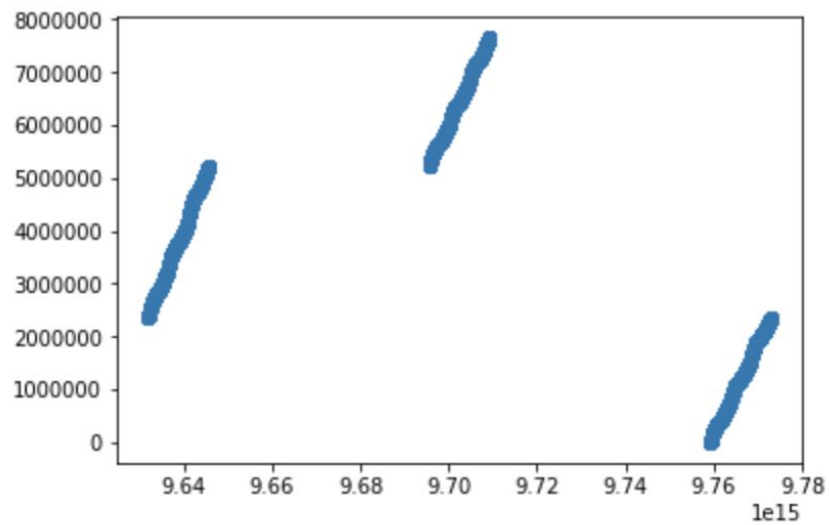
# Appendix

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# Dealing with time obfuscation

- Time obfuscation maintains monotonicity
- Use smallest possible time delta from given numbers
- Quadrillion ( $10^{15}$ )





Time column EDA

# Class imbalance

- 1910 Classified Humans
- 103 Classified Bots
- Randomly undersampling real  
bidders before modeling