Work summary

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The methods used in this work and their explanations:

• Models:

'Weak' learners:

- KNN
- Logistic Regression
- Decision Tree

Ensemble learning:

- Random Forest: estimate the result of a collection of decision trees.
- Voting Classification:

Hard: aggregate the predictions of each classifier and predict the class that gets the most votes.

Soft: predict the class with the highest-class probability, averaged over all the individual classifiers.

- Stacking: train a model to perform the aggregation of voting.

Boosting:

- <u>Adaboosting:</u> each successor model fits according to the weight of its predecessor's mistakes.
- <u>Gradient Boosting:</u> same as adaboost, but this method tries to fit the new predictor to the residual errors made by the previous predictor.
- XGBoosting: Extreme Gradient Boosting

Clustering:

- K-means

• Dimensionality Reduction:

- <u>PCA</u>: choosing a lower dimensional hyperplane of the data that preserve X value of the original data's variance. (Number of features that preserve that variance)

Classification Enhancements:

- Random Forest, Voting, Stacking
- Adaboost, XGBoost

FMNIST:

- KNN, Logistic Regression, Decision Tree
- Random Forest, Voting, Stacking
- Adaboost, XGBoost
- PCA

Cats vs Dogs:

- KNN, Logistic Regression, Decision Tree
- Random Forest, Voting, Stacking
- Adaboost, XGBoost
- K-means
- PCA

<u>Hands:</u>

- KNN, Logistic Regression, Decision Tree
- Random Forest, Voting, Stacking
- Adaboost, XGBoost
- K-means
- PCA