

LIMEaid Local Interpretable Model-agnostic Explanations (LIME)

Data 515, Spring 2019 M.S. Data Science

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Interpretability in Machine Learning

Some highly accurate models are not "explainable"

Neural networks, random forests

Why is this a problem?

- Bias, not obvious
- High test set accuracy but poor results in the field
- Policy or law demands an explanation of any decision

Solution: model-agnostic local explanations

- Explain one instance, not entire model
- Fit a simple model to explain a small section of decision space

LIMEaid: A LIME solution for tabular data

Input

- A "complex" ML model, fit by sklearn classifier object with .predict
- An instance of data (x) and its model output (f(x))
- Probability domain for normalized predictor variables (histograms)

Output

- Sparse linear models (few features), plottable
- List significant features

Install

PyPI package

Use Cases

User profile: data scientist with Python programming experience

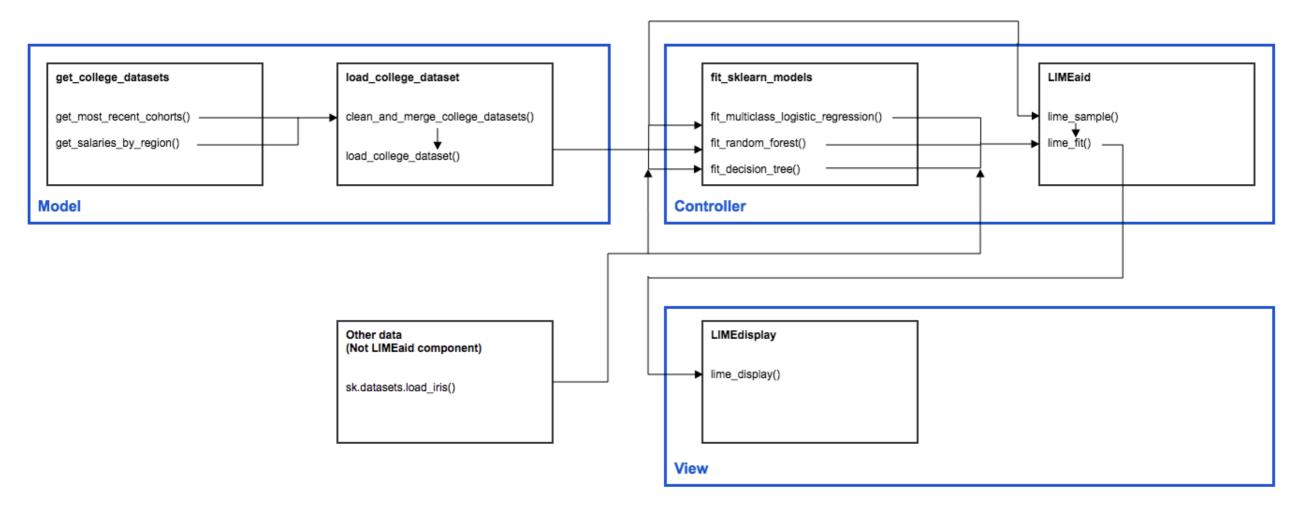
1. Model verification scenario

- User wants to preempt poor model performance "in the field"
- Use LIMEaid to sample test dataset
- Show most significant features for decision
- Tune or replace model if spurious correlation or other issues

2. Decision explanation scenario

- Classification has already been made by a model
- Use LIMEaid to sample whole dataset
- Produce easy-to-explain "two-dimensional" plot of linear regression

Design



Data

Sources

- College Scorecard (data.gov): Annual report of schools and attributes (SAT scores, majors offered, region, cost, public/private/for profit, etc.)
- "Where it Pays to Attend College" (<u>Kaggle.com</u>) obtained from (<u>Wall Street Journal</u>), based on Payscale, Inc. (<u>College Salary Report Methodology</u>): Article reporting salaries of graduates, salaries by major, etc.

Merge

- Significant cleaning, reformatting to match sets on college name
- String manipulation, removal of hyphens, abbreviations, region names, etc.

More

Sklearn's provided "Iris" data

Models

Scikit-learn classifiers that predict probabilities (predict_proba implemented)

Multiclass logistic regression (sklearn.linear model.LogisticRegression)

85% accuracy on College data

Random Forest (sklearn.ensemble.randomforestclassifier)

65% accuracy on College data

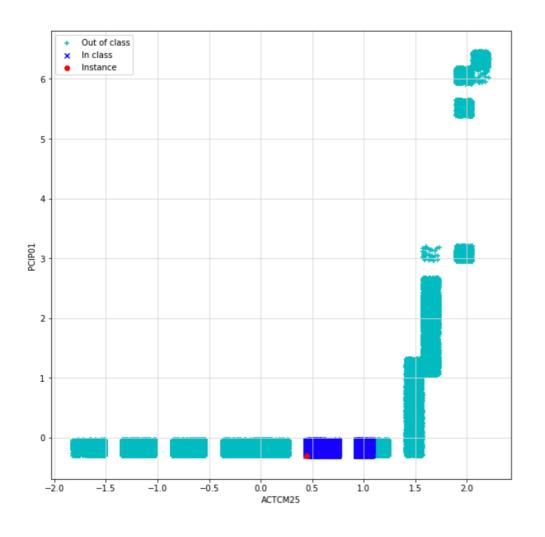
Decision tree (sklearn.tree)

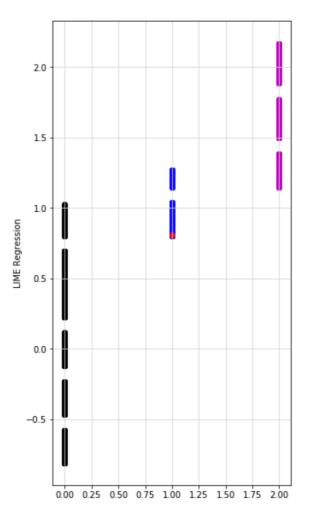
65% accuracy on College data

Models not tuned for improved accuracy (default settings)

Demo

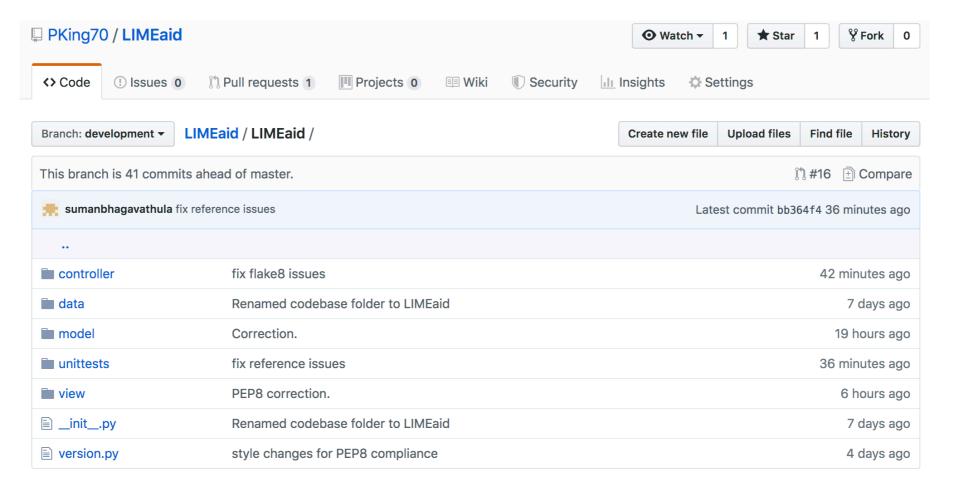
Education and Iris data





Project Structure

- LIMEaid Github Repo based on Shablona
- Used Model-View-Controller (MVC) codebase architecture



Future Work

API support for data acquisition to support dynamic features:

 College Scorecard (data.gov) currently published with new features and data dictionary yearly

Model tuning for examples:

• Currently using defaults, could improve accuracy > 65%

Modify penalty for number of coefficients

Improve method of generating random samples

More data type and model support: image data, NLP support, support for model objects beyond sklearn classifiers

Questions?