Solutions of Ex.1

27.03.2017

- 1. Construct a classification pipeline
- 2. Normalize text
- 3. Extract features from text
- 4. Experiment with classifiers
- 5. Evaluate results

Constructing a Pipeline : pipeline.py

- 1. Read the data
- 2. Split data for cross-validation
- 3. For each fold:
 - 3.1. Extract features
 - 3.2. Run Classifier
 - 3.3. Evaluate Classifier

1. Reading Data: get_dataset

- Use FNC's implementation of load_dataset
- 2. Link Body Ids to Body text (in separate file)
- 3. Add dictionary, which will hold the features

Instance:

```
{'Body ID': '712',
    'Headline': "Police find mass graves with
at least '15 bodies' near Mexico "
        'town where 43 students disappeared
after police clash',
    'Stance': 'unrelated',
    'articleBody': 'Danny Boyle is directing the
untitled film\n ... later was responsible for
creating the early Apple computers.',
    'features': {}
}
```

2. Cross-Validation: test_cross_validation

- Get an instance of KFold from sklearn.model_selection
- 2. **Apply** k-fold over the dataset and **Iterate** over the folds
- 3. For each fold extract features, train classifier, predict labels, evaluate labels
- 4. Get average score for the folds

```
k_folds = KFold(n_splits=2, random_state=42)
for train index, test index in k_folds.split(X):
    X train = [X[i] for i in train index]
    X test = [X[i] for i in test index]
    predicted labels =
run_classifier(train=X_train, test=X_test)
    score = get_fnc_score(X test,
predicted labels)
```

3. For each fold: run_classifier + get_fnc_score

- 1. Use sklearn's **Pipeline** to extract all features (put them in the features dictionary of each instance).
- 2. Include the **classifier** in the pipeline or run it separately
- 3. **Predict** the classes for the test data, running it over the same trained Pipeline

```
pipeline = Pipeline([
('preprocess_lemmas', TokenizedLemmas()),
('sent len', SentenceLength()),
('tfidf', BagOfTfIDF(train)),
('pos', POS()),
('ner', NER()),
('word overlap', WordOverlap()),
('transform', ToMatrix()),
('norm', MinMaxScaler()),
('clf', SVC())])
pipeline.fit(train, true labels)
pipeline.predict(test)
```

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Normalize Text: preprocess.py

TokenizedLemmas: get list of tokenized lemmas in body and headline. This can be later, in the pipeline, reused by the various features.

Different Normalizations of Text:

- _remove_stopwords
- _normalize_word
- _get_tokenized_lemmas
- _clean lowercasing, removing non-alphanumerics

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3. Extract Features from text: features.py

class Feature(TransformerMixin):

"""Feature Interface."""

def fit(self, X, y=None, **fit_params):

return self

Features in the sklearn's Pipeline should implement **TransformerMixin**

2 abstract methods:

- fit (train-not always needed
 => abstract class Feature)
- 2. transform (produce features)

3. Extract Features from text: features.py

- 1. Subclass Feature for the separate features and implement at least transform method.
- 2. Add each feature to the dictionary of features.
- 3. Use ToMatrix to turn dictionary to NxM matrix of features with N instances and M features.
- 4. Normalize features to the [0,1] scale.

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4. Experiment with classifiers

```
class sklearn.svm.SVC(C=1.0, kernel='rbf', degree=3, gamma='auto', coef0=0.0, shrinking=True, probability=False, tol=0.001, cache_size=200, class_weight=None, verbose=False, max_iter=-1, decision_function_shape=None, random_state=None)[source]¶
```

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5. Evaluate Results

- get_fnc_score modified accuracy score
- sklearn.metrics:
 - accuracy_score
 - precision_score
 - classification report
 - recall_score
 - f1_score
 - o confision_matrix