

Pipelines, feature & text preprocessing

CASE STUDY: SCHOOL BUDGETING WITH MACHINE LEARNING IN PYTHON



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The pipeline workflow

- Repeatable way to go from raw data to trained model
- Pipeline object takes sequential list of steps
 - Output of one step is input to next step
- Each step is a tuple with two elements
 - Name: string
 - Transform: obj implementing `.fit()` and `.transform()`
- Flexible: a step can itself be another pipeline!

Instantiate simple pipeline with one step

```
from sklearn.pipeline import Pipeline
from sklearn.linear_model import LogisticRegression
from sklearn.multiclass import OneVsRestClassifier
```

```
pl = Pipeline([
    ('clf', OneVsRestClassifier(LogisticRegression()))
])
```

Train and test with sample numeric data

```
sample_df.head()
```

	label	numeric	text	with_missing
0	a	-4.167578	bar	-4.084883
1	a	-0.562668		2.043464
2	a	-21.361961		-33.315334
3	b	16.402708	foo bar	30.884604
4	a	-17.934356	foo	-27.488405

Train and test with sample numeric data

```
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(
    sample_df[['numeric']],
    pd.get_dummies(sample_df['label']),
    random_state=2)

pl.fit(X_train, y_train)
```

```
Pipeline(steps=[('clf', OneVsRestClassifier(estimator=LogisticRegression(C=1.0,
class_weight=None, dual=False, fit_intercept=True,
    intercept_scaling=1, max_iter=100, multi_class='ovr', n_jobs=1,
    penalty='l2', random_state=None, solver='liblinear', tol=0.0001,
    verbose=0, warm_start=False),
    n_jobs=1))])
```

Train and test with sample numeric data

```
accuracy = pl.score(X_test, y_test)
print('accuracy on numeric data, no nans: ', accuracy)
```

```
accuracy on numeric data, no nans: 0.44
```

Adding more steps to the pipeline

```
X_train, X_test, y_train, y_test = train_test_split(sample_df[['numeric',  
                                                         'with_missing']], pd.get_dummies(  
                                                         sample_df['label']), random_state=2)  
  
pl.fit(X_train, y_train)
```

Traceback (most recent call last):

...

ValueError: Input contains NaN, infinity or a value too large for
dtype('float64').

Preprocessing numeric features with missing data

```
from sklearn.preprocessing import Imputer
X_train, X_test, y_train, y_test = train_test_split(
    sample_df[['numeric', 'with_missing']],
    pd.get_dummies(sample_df['label']),
    random_state=2)

pl = Pipeline([
    ('imp', Imputer()),
    ('clf', OneVsRestClassifier(LogisticRegression()))
])
```


Preprocessing numeric features with missing data

```
pipeline.fit(X_train, y_train)
accuracy = pl.score(X_test, y_test)
print('accuracy on all numeric, incl nans: ', accuracy)
```

```
accuracy on all numeric, incl nans: 0.48
```

- No errors!

Let's practice!

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Text features and feature unions

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Preprocessing text features

```
from sklearn.feature_extraction.text import CountVectorizer
X_train, X_test, y_train, y_test = train_test_split(sample_df['text'],
                                                    pd.get_dummies(
                                                        sample_df['label']),
                                                    random_state=2)

pl = Pipeline([
    ('vec', CountVectorizer()),
    ('clf', OneVsRestClassifier(LogisticRegression()))
])
```

Preprocessing text features

```
pl.fit(X_train, y_train)
```

```
Pipeline(steps=[('vec', CountVectorizer(analyzer='word', binary=False,
decode_error='strict', dtype=<class 'numpy.int64'>, encoding='utf-8',
input='content', lowercase=True, max_df=1.0, max_features=None, min_df=1,
ngram_range=(1, 1), preprocessor=None, stop_words=None, strip_...=None,
solver='liblinear', tol=0.0001, verbose=0, warm_start=False), n_jobs=1))])
```

```
accuracy = pl.score(X_test, y_test)
print('accuracy on sample data: ', accuracy)
```

```
accuracy on sample data: 0.64
```

Preprocessing multiple dtypes

- Want to use **all** available features in one pipeline
- Problem
 - Pipeline steps for numeric and text preprocessing can't follow each other
 - e.g., output of `CountVectorizer` can't be input to `Imputer`
- Solution
 - `FunctionTransformer()` & `FeatureUnion()`

FunctionTransformer

- Turns a Python function into an object that a scikit-learn pipeline can understand
- Need to write two functions for pipeline preprocessing
 - Take entire DataFrame, return numeric columns
 - Take entire DataFrame, return text columns
- Can then preprocess numeric and text data in separate pipelines

Putting it all together

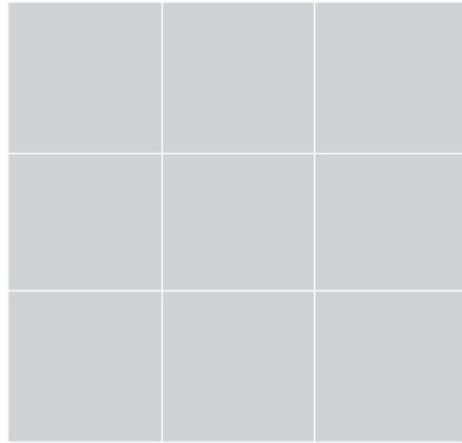
```
X_train, X_test, y_train, y_test = train_test_split(sample_df[['numeric',  
                                                             'with_missing', 'text']], pd.get_dummies(  
                                                             sample_df['label']), random_state=2)  
  
from sklearn.preprocessing import FunctionTransformer  
from sklearn.pipeline import FeatureUnion
```


Putting it all together

```
get_text_data = FunctionTransformer(lambda x: x['text'],  
                                     validate=False)  
  
get_numeric_data = FunctionTransformer(lambda x: x[['numeric',  
                                                    'with_missing']], validate=False)
```

FeatureUnion Text and Numeric Features

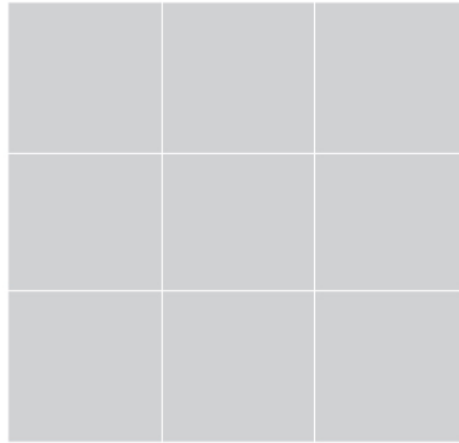
Text Features



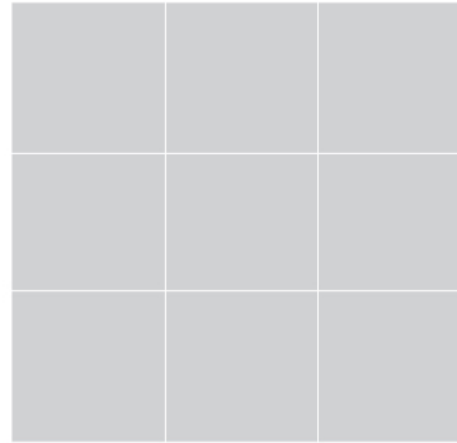
```
from sklearn.pipeline import FeatureUnion
```

FeatureUnion Text and Numeric Features

Text Features



Numeric Features

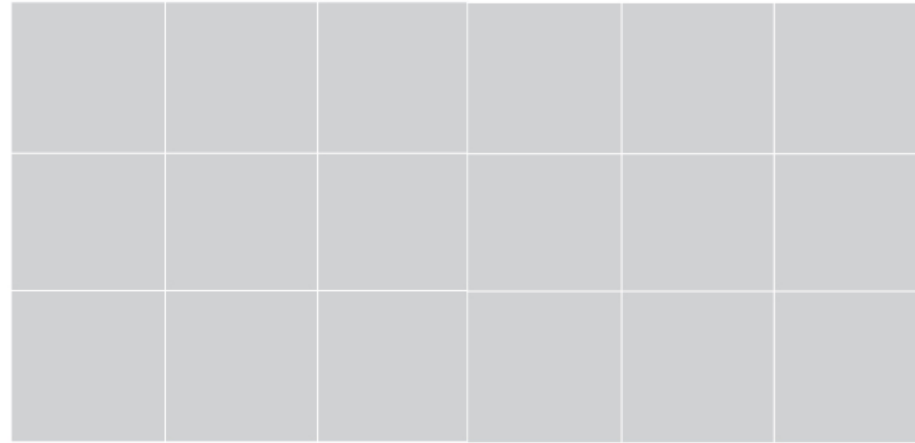


```
from sklearn.pipeline import FeatureUnion
```

FeatureUnion Text and Numeric Features

Text Features

Numeric Features



```
from sklearn.pipeline import FeatureUnion
union = FeatureUnion([
    ('numeric', numeric_pipeline),
    ('text', text_pipeline)
])
```

Putting it all together

```
numeric_pipeline = Pipeline([
    ('selector', get_numeric_data),
    ('imputer', Imputer())
])

text_pipeline = Pipeline([
    ('selector', get_text_data),
    ('vectorizer', CountVectorizer())
])

pl = Pipeline([
    ('union', FeatureUnion([
        ('numeric', numeric_pipeline),
        ('text', text_pipeline)
    ])),
    ('clf', OneVsRestClassifier(LogisticRegression()))
])
```

Let's practice!

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Choosing a classification model

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Main dataset: lots of text

```
LABELS = ['Function', 'Use', 'Sharing', 'Reporting', 'Student_Type',  
          'Position_Type', 'Object_Type', 'Pre_K', 'Operating_Status']  
NON_LABELS = [c for c in df.columns if c not in LABELS]  
len(NON_LABELS) - len(NUMERIC_COLUMNS)
```

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Using pipeline with the main dataset

```
import numpy as np
import pandas as pd
df = pd.read_csv('TrainingSetSample.csv', index_col=0)
dummy_labels = pd.get_dummies(df[LABELS])
X_train, X_test, y_train, y_test = multilabel_train_test_split(
    df[NON_LABELS], dummy_labels,
    0.2)
```

Using pipeline with the main dataset

```
get_text_data = FunctionTransformer(combine_text_columns,  
                                   validate=False)  
  
get_numeric_data = FunctionTransformer(lambda x:  
                                     x[NUMERIC_COLUMNS], validate=False)  
  
pl = Pipeline([  
    ('union', FeatureUnion([  
        ('numeric_features', Pipeline([  
            ('selector', get_numeric_data),  
            ('imputer', Imputer())  
        ])),  
        ('text_features', Pipeline([  
            ('selector', get_text_data),  
            ('vectorizer', CountVectorizer())  
        ]))  
    ]),  
    ('clf', OneVsRestClassifier(LogisticRegression()))  
])
```

Performance using main dataset

```
pl.fit(X_train, y_train)
```

```
Pipeline(steps=[('union', FeatureUnion(n_jobs=1,  
    transformer_list=[('numeric_features', Pipeline(steps=  
    [('selector', FunctionTransformer(accept_sparse=False,  
    func=<function <lambda> at 0x11415ec80>, pass_y=False,  
    validate=False)), ('imputer', Imputer(axis=0, copy=True,  
    missing_valu...=None, solver='liblinear', tol=0.0001,  
    verbose=0, warm_start=False),n_jobs=1))])])])
```

Flexibility of model step

- Is current model the best?
- Can quickly try different models with pipelines
 - Pipeline preprocessing steps unchanged
 - Edit the model step in your pipeline
 - Random Forest, Naïve Bayes, k-NN

Easily try new models using pipeline

```
from sklearn.ensemble import RandomForestClassifier
pl = Pipeline([
    ('union', FeatureUnion(
        transformer_list = [
            ('numeric_features', Pipeline([
                ('selector', get_numeric_data),
                ('imputer', Imputer())
            ])),
            ('text_features', Pipeline([
                ('selector', get_text_data),
                ('vectorizer', CountVectorizer())
            ]))
        ]
    )),
    ('clf', OneVsRest(RandomForestClassifier()))
])
```

Let's practice!

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