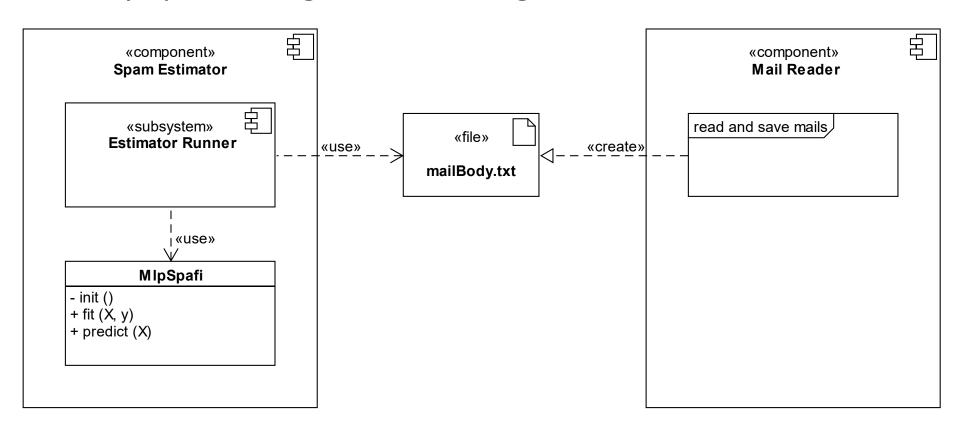
# Ham | | Spam

Email Spam Estimator
w/ TensorFlow and UiPath
DH

### .:: Overview

#### Task: Identify spam using email messages



### .:: Under The Hood: Spam Estimator / DE

- csv text file with 5728 emails, each labelled as spam (1) or ham (0)
- samples
  - start with 'Subject: '
  - lower case text
  - contain words, numbers, non-word chars
  - min: 13 chars, max: 43952 chars
  - English only
- unbalanced, sorted dataset
  - 1368 spam, 4360 ham

Subject: contact info i will be in one of these two places -- my home: 011 91 80 3312635 my in - laws ' home: 011 91 80 5262719 you can also contact me by email at vshanbh @ yahoo. com, but it is better to call since i do not have easy access to a computer, and there may be a delay with reading email. vasant, 0

# .:: Under The Hood: Spam Estimator / DP

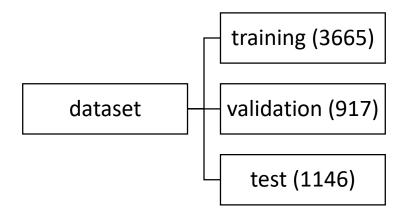
- remove string 'Subject: '
- convert text documents to a matrix of token counts

#### Example:

```
text_docs =
['contact info i will be in one of
these two places',
'but it is better to call since i
do not have']

['better',
[[0. 1. 1. 1.] 'contact',
[1. 0. 0. 0.]] 'info', 'places']
```

- split and shuffle dataset
  - 20% for test
  - 20% of remaining set for validation
  - balance of spam & ham persists



# .:: Under The Hood: Spam Estimator / ME

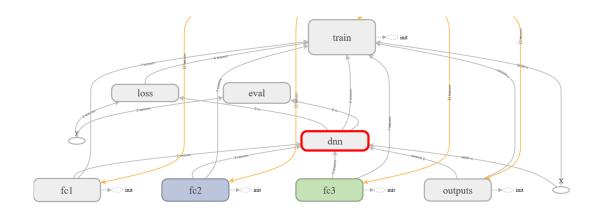
Question: Is it better to identify spam as ham or ham as spam?

#### **Confusion Matrix:**

|            |          | predicted class |         |
|------------|----------|-----------------|---------|
|            |          | spam (1)        | ham (0) |
| true class | spam (1) | TP              | FN      |
|            | ham (0)  | FP              | TN      |

Metric to use is Precision:  $PPV = \frac{TP}{TP+FP}$ 

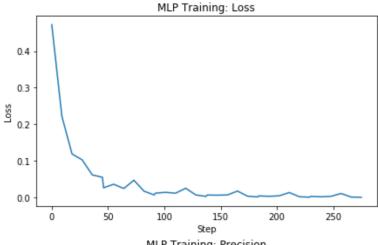
### .:: Under The Hood: Spam Estimator / IM

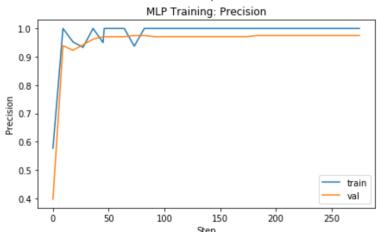


epoch # 0, batch # 46 / 46, train metric: 0.95 , val metric: 0.9704641350210971 val metric improved, model saved to ./tf\_saves/run\_MlpSpafi/best\_acc\_val.ckpt epoch # 1, batch # 46 / 46, train metric: 1.0 , val metric: 0.9710743801652892 val metric improved, model saved to ./tf\_saves/run\_MlpSpafi/best\_acc\_val.ckpt epoch # 2, batch # 46 / 46, train metric: 1.0 , val metric: 0.9710743801652892 epoch # 3, batch # 46 / 46, train metric: 1.0 , val metric: 0.975103734439834 val metric improved, model saved to ./tf\_saves/run\_MlpSpafi/best\_acc\_val.ckpt epoch # 4, batch # 46 / 46, train metric: 1.0 , val metric: 0.975103734439834 epoch # 5, batch # 46 / 46, train metric: 1.0 , val metric: 0.975103734439834 early stopping after epoch 5

# .:: Under The Hood: Spam Estimator / EV

#### **Training**





#### Scores

validation: 0.9751

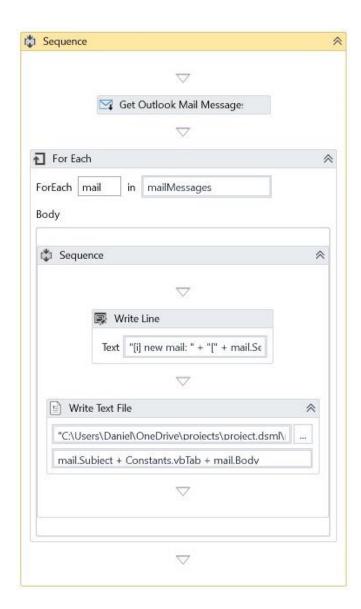
test: 0.9893

#### **Confusion Matrix:**

#### predicted class

|            |          | spam (1) | ham (0) |
|------------|----------|----------|---------|
| true class | spam (1) | 280      | 10      |
|            | ham (0)  | 3        | 853     |

# .:: Under The Hood: Mail Reader / SEQ



#### .:: Real-Time Demonstration

- Python script: Spam Estimator
- UiPath sequence: Mail Reader
- choose and send mails
  - http://www.antespam.co.uk/spam-resource/
  - mailto:dan.spamcheck@outlook.com

### .:: Improvements

- data preprocessing
  - lemmatization / stemming
  - correlations, i.e. mail length vs spam
- implementation of estimator in UiPath
  - Python + required packages (w/o virtual environment)
  - official package 'UiPath.Python.Activities'
- serving mode of estimator and online learning capability
- check of other mail types and languages
  - attachments
  - graphical mails

### .:: Interesting Facts

- underlying MLP was prior used to recognize handwritten digits
- dataset contains mails from the Enron scandal
- less top k words worked better