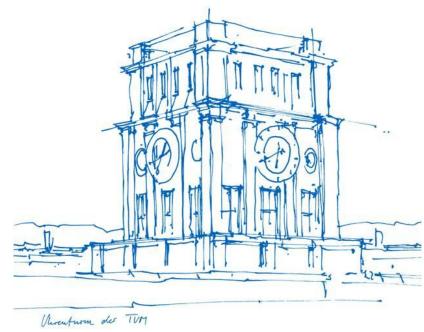


Comparing **Sentiment Analysis** of Discourse Units and Sentences

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Progress Report

- Preprocessing code finished.
 - lemmatize
 - stopwords remove
 - length padding
- MilNet model running.
 - XLING feature based
 - word2vec feature based
- → Memory usage not suitable for Colab
- → Model accuracy not satisfiable

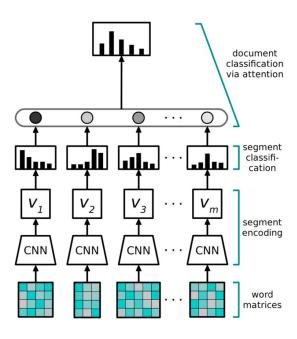


Embedding Layer

(None, 20, 30, 300) -> 272.8 GB in total!

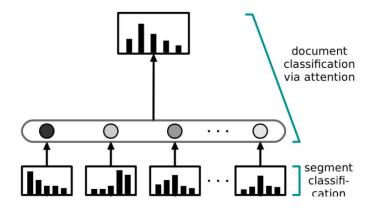
```
embedding_layer = tf.keras.layers.Embedding(
    input_dim=w2v.shape[0],
    output_dim=w2v_len,
    weights=[w2v],
    input_length=max_word,
    trainable=False
) (model_input)
```

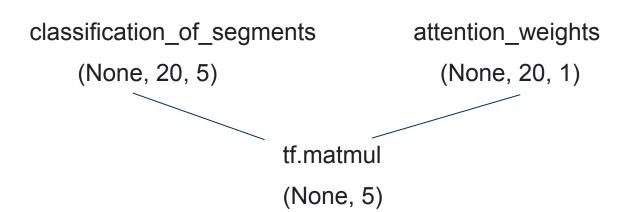
(None, 20, 30) -> Less than 2 GB.





Parameter not Found?







Parameter not Found?

```
classification_model = ...
attention_model = ...
```

Option 1: tf.matmul

```
weighted_layer = tf.keras.layers.Lambda(tf.matmul,
                      arguments={'transpose_a': True,
                                   'b':attention_model})(classification_model)
squeeze_layer = tf.keras.layers.Lambda(tf.squeeze, arguments={'axis':
-1})(weighted_layer)
```

Option 2: tf.keras.layers.Multiply

```
weighted_layer = tf.keras.layers.Multiply()([attention_model, classification_model])
reduce_layer = tf.keras.layers.Lambda(tf.reduce_mean, arguments={'axis':1})
                                       (weighted_layer)
```



Parameter not Found?

Constructing Model ...
Model Constructed. Compiling ...
Model Compiled.
Model: "model_95"

Layer (type)	Output Shape	Param #	
input_73 (InputLayer)	[(None, 20, 30)]	0	
embedding_3 (Embedding)	(None, 20, 30, 300)	58749600	
model_92 (Model)	(None, 20, 210)	5061000	
model_94 (Model)	(None, 20, 5)	21100	
lambda_404 (Lambda)	(None, 5, 1)	0	
lambda 405 (Lambda)	(None, 5)	0	

Total params: 63,831,700 Trainable params: 5,073,700 Non-trainable params: 58,758,000 Constructing Model ...
Model Constructed. Compiling ...
Model Compiled.
Model: "model_47"

Layer (type)	Output S	hape	Param #	Connected to
input_25 (InputLayer)	[(None,	20, 30)]	0	
embedding_1 (Embedding)	(None, 2	0, 30, 300)	58749600	input_25[0][0]
model_44 (Model)	(None, 2	0, 210)	5061000	embedding_1[0][0]
sequential_161 (Sequential)	(None, 2	0, 300)	325800	model_44[1][0]
model_45 (Model)	(None, 2	0, 1)	1812000	sequential_161[0][0]
model_46 (Model)	(None, 2	0, 5)	21100	model_44[1][0]
multiply_1 (Multiply)	(None, 2	0, 5)	0	model_45[1][0] model_46[1][0]
lambda_201 (Lambda)	(None, 5)	0	multiply_1[0][0]

Total params: 65,969,500 Trainable params: 7,211,500 Non-trainable params: 58,758,000



SparseCategoricalEntropy

```
model.compile(
[...]
loss=tf.keras.losses.CategoricalCrossentropy(),
[...]
)
ValueError: You are passing a target array of shape (1000, 1) while using as loss
`categorical_crossentropy`. `categorical_crossentropy` expects targets to be binary matrices (1s and 0s) of shape (samples, classes)
```

→ from keras.utils import to_categorical

```
y_binary = to_categorical(y_int)
```

Alternatively, you can use the loss function `sparse_categorical_crossentropy` instead, which does expect integer targets.



SparseCategoricalEntropy

tf.keras.losses.SparseCategoricalEntropy

- Available after version 1.14-rc
- Compute the categorical cross entropy between feature and integer labels

```
cce = tf.keras.losses.SparseCategoricalCrossentropy()
loss = cce(
[0, 1, 2],
[[.9, .05, .05], [.5, .89, .6], [.05, .01, .94]]) # Loss: 0.3239
```

tf.keras.losses.CategoricalEntropy

- Compute the categorical cross entropy between feature and one-hot labels
- Compute the categorical cross entropy between feature and integer labels (before version 1.14-rc)

```
cce = tf.keras.losses.CategoricalCrossentropy()
loss = cce(
[[1., 0., 0.], [0., 1., 0.], [0., 0., 1.]],
[[.9, .05, .05], [.5, .89, .6], [.05, .01, .94]]) # Loss: 0.3239
```



Solved Problems

- Appending to np.array too slow: Do not use np.append frequently, use
 np.zeros to allocate the array first
- Dumping lots of npy files requires lots of time: Use hdf5 file to store bunch
 of np.array into a single file
- Cannot load dataset at once: Use fit generator instead of fit.

Still Existing Problems

Model accuracy: Barely higher than 20% for word2vec features



XLING Feature Based Model

- On 1000 reviews
- 5 Balanced classes
- Accuracy:

Validation: 75%

Macro F1-Score: 0.67

Next: Integrate XLING into Dataloader

→ Train on whole dataset



Next Steps

- Keep debugging the model with small dataset
- Train on full dataset with sentence/EDU features
- Integrate data loading / embeddings process into a continuous df.data pipeline
- Use different segment embeddings e.g. BERT, ELMo



Muchas Gracias!

Li Canchen, Hendrik Pauthner, Tim Pfeifle Munich, 22. May 2019

