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# “Assessing Humor in Edited News Headlines”

introduced by *Vinh N. & Finn R.*

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# Given Dataset & Task

dev

id	original	edit	grades	meanGrade
1723	Thousands of gay and bisexual <men/> convicted of long-abolished sexual offences are posthumously pardoned	swans	22100	1.0
12736	Special <prosecutor/> appointed to Trump Russia	chef	21100	0.8
12274	Spanish police detain man and search Ripoll addresses in hunt for terror <suspects/>	squad	21000	0.6
8823	N.Y. Times <reprimands/> reporter for sharing ' unfounded rumor ' about Melania Trump	applauds	32210	1.6

- Specs
  - 12071 datapoints
  - 0.0 is min(grades)
  - 3.0 is max(grades)

# 3 Different Approaches

**Approach-1**

[ X ]

**Approach-2**

[ ]

**Approach-3**

[ ]

# Given Dataset & Interpretation

dev

id	original	edit	grades	meanGrade
1723	Thousands of gay and bisexual <men/> convicted of long-abolished sexual offences are posthumously pardoned	swans	22100	1.0
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[“Special **prosecutor** appointed to Trump Russia”] → [ 0.0 ]

[“Special **chef** appointed to Trump Russia”] → [ 0.8 ]

# Preprocessing Dataset

## I. Unfold dataset (accordingly to interpretation)

- A. [ "Special **prosecutor** appointed to Trump Russia" ] → [ 0.0 ]
- B. [ "Special **chef** appointed to Trump Russia" ] → [ 0.8 ]

## II. Stemming/Lemmatize & Tokenize according to word frequency

- A. [ 12, 833, 401, 6, 7, 9, ] → [ 0.0 ]
- B. [ 12, 1050, 401, 6, 7, 9, ] → [ 0.8 ]

## III. Padding sequences

- A. [ 0, 0, 0, 12, 833, 401, 6, 7, 9, ] → [ 0.0 ]

## IV. Normalize dataset

- A. [ 0, 0, 0, 12/n, 833/n, 401/n, 6/n, 7/n, 9/n, ] → [ 0.0/m ]

# Network -

```
25 # Parameters
1  embed_dim = 25
2  batch_size = 32
3  epochs = 25
4  learning_rate = 0.001
5
6  # Model
7  model = Sequential()
8  model.add(Dense(embed_dim, input_dim=embed_dim, activation='relu'))
9  model.add(Dense(200, activation='relu'))
10 model.add(Dense(200, activation='relu'))
11 model.add(Dense(1, activation='sigmoid'))
12 opt = Optimizers.Adam(lr=learning_rate)
13 model.compile(loss='mse', optimizer=opt, metrics=['accuracy'])
14
```

# Results

```
Epoch 22/25  
19313/19313 [=====] - 1s 30us/step - loss: 0.0408 - accuracy: 0.5271 - val_loss: 0.0434 - val_accuracy: 0.5256  
Epoch 23/25  
19313/19313 [=====] - 1s 32us/step - loss: 0.0407 - accuracy: 0.5271 - val_loss: 0.0434 - val_accuracy: 0.5256  
Epoch 24/25  
19313/19313 [=====] - 1s 31us/step - loss: 0.0406 - accuracy: 0.5271 - val_loss: 0.0441 - val_accuracy: 0.5256  
Epoch 25/25  
19313/19313 [=====] - 1s 31us/step - loss: 0.0405 - accuracy: 0.5271 - val_loss: 0.0432 - val_accuracy: 0.5256  
4829/4829 [=====] - 0s 10us/step  
Test loss: 0.04317193100802472  
Test accuracy: 0.5255746245384216  
Persist model completely in 'models/fcc_v1'.  
(venv) potato:approach1 vngu$
```

mean squared error!

# Given Dataset & Interpretation

dev

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1723	Thousands of gay and bisexual <men/> convicted of long-abolished sexual offences are posthumously pardoned	swans	22100	1.0
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8823	N.Y. Times <reprimands/> reporter for sharing ' unfounded rumor ' about Melania Trump	applauds	32210	1.6

~~[“Special prosecutor appointed to Trump Russia”]~~ → ~~[0.0]~~  
[“Special chef appointed to Trump Russia”] → [ 0.8 ]



# Results

```
9656/9656 [=====] - 0s 31us/step - loss: 0.0231 - accuracy: 0.0543 - val_loss: 0.0486 - val_accuracy: 0.0480
Epoch 97/100
9656/9656 [=====] - 0s 31us/step - loss: 0.0231 - accuracy: 0.0543 - val_loss: 0.0480 - val_accuracy: 0.0497
Epoch 98/100
9656/9656 [=====] - 0s 31us/step - loss: 0.0228 - accuracy: 0.0543 - val_loss: 0.0465 - val_accuracy: 0.0501
Epoch 99/100
9656/9656 [=====] - 0s 32us/step - loss: 0.0229 - accuracy: 0.0543 - val_loss: 0.0485 - val_accuracy: 0.0484
Epoch 100/100
9656/9656 [=====] - 0s 31us/step - loss: 0.0227 - accuracy: 0.0543 - val_loss: 0.0486 - val_accuracy: 0.0489
2415/2415 [=====] - 0s 10us/step
Training loss: 0.04855791383401701
Training accuracy: 0.04886128380894661
Persist model completely in 'models/fcc_v1'.
(venv) potato:approach1-1 vngu$
```

mean squared error!

# 3 Different Approaches

**Approach-1**

[ X ]

**Approach-2**

[ X ]

**Approach-3**

[ ]

# Given Dataset & Interpretation

dev

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[“Special **prosecutor** appointed to Trump Russia”] → [“0.0”]

[“Special **chef** appointed to Trump Russia”] → [“0.8”]

**31 different grades  
means 31 diff. classes!**

# Preprocessing Dataset

## I. Unfold dataset (accordingly to interpretation)

- A. [ "Special **prosecutor** appointed to Trump Russia" ] → [ "0.0" ]
- B. [ "Special **chef** appointed to Trump Russia" ] → [ "0.8" ]

## II. Stemming/Lemmatize & Tokenize according to word frequency

- A. [ 12, 833, 401, 6, 7, 9, ] → [ "0.0" ]
- B. [ 12, 1050, 401, 6, 7, 9, ] → [ "0.8" ]

## III. Padding sequences

- A. [ 0, 0, 0, 12, 833, 401, 6, 7, 9, ] → [ "0.0" ]

## IV. Normalize dataset

- A. [ 0, 0, 0, 12/n, 833/n, 401/n, 6/n, 7/n, 9/n, ] → [ "0.0" ]

# Network -

```
17 # Parameters
1  embed_dim = 25
2  batch_size = 32
3  epochs = 100
4  num_classes = 31
5
6  # Model
7  model = Sequential()
8  model.add(Dense(256, activation='relu', input_dim=embed_dim))
9  model.add(Dropout(0.5))
10 model.add(Dense(512, activation='relu', input_dim=embed_dim))
11 model.add(Dropout(0.5))
12 model.add(Dense(num_classes, activation='softmax'))
13 model.compile(loss='categorical_crossentropy', optimizer='adam', metrics=['accuracy'])
14 print(model.metrics_names)
15
```

# Results

```
Epoch 96/100
17381/17381 [=====] - 1s 56us/step - loss: 1.7021 - accuracy: 0.5335 - val_loss: 1.8350 - val_accuracy: 0.5274
Epoch 97/100
17381/17381 [=====] - 1s 56us/step - loss: 1.7061 - accuracy: 0.5337 - val_loss: 1.8358 - val_accuracy: 0.5264
Epoch 98/100
17381/17381 [=====] - 1s 57us/step - loss: 1.7033 - accuracy: 0.5347 - val_loss: 1.8412 - val_accuracy: 0.5254
Epoch 99/100
17381/17381 [=====] - 1s 57us/step - loss: 1.7011 - accuracy: 0.5341 - val_loss: 1.8379 - val_accuracy: 0.5274
Epoch 100/100
17381/17381 [=====] - 1s 56us/step - loss: 1.7011 - accuracy: 0.5320 - val_loss: 1.8323 - val_accuracy: 0.5274
4829/4829 [=====] - 0s 13us/step
Test loss: 1.9399894609488817
Test accuracy: 0.5141851305961609
Persist model completely in 'models/fcc_v1'.
(venv) potato:approach2 vngu$
```

mean squared error!

# Given Dataset & Interpretation

dev

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1723	Thousands of gay and bisexual <men/> convicted of long-abolished sexual offences are posthumously pardoned	swans	22100	1.0
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[~~"Special prosecutor appointed to Trump Russia"~~]

→

[~~"0.0"~~]

["Special chef appointed to Trump Russia"]

→

["0.8"]

31 different grades  
means 31 diff. classes!

# Results

```
8690/8690 [=====] - 0s 56us/step - loss: 2.2917 - accuracy: 0.1975 - val_loss: 2.5283 - val_accuracy: 0.1201
Epoch 96/100
8690/8690 [=====] - 0s 56us/step - loss: 2.2903 - accuracy: 0.1952 - val_loss: 2.5298 - val_accuracy: 0.1263
Epoch 97/100
8690/8690 [=====] - 0s 56us/step - loss: 2.2929 - accuracy: 0.1959 - val_loss: 2.5334 - val_accuracy: 0.1170
Epoch 98/100
8690/8690 [=====] - 0s 56us/step - loss: 2.2904 - accuracy: 0.1992 - val_loss: 2.5376 - val_accuracy: 0.1284
Epoch 99/100
8690/8690 [=====] - 0s 56us/step - loss: 2.2880 - accuracy: 0.1972 - val_loss: 2.5328 - val_accuracy: 0.1335
Epoch 100/100
8690/8690 [=====] - 0s 56us/step - loss: 2.2741 - accuracy: 0.2079 - val_loss: 2.5405 - val_accuracy: 0.1284
2415/2415 [=====] - 0s 14us/step
Test loss: 2.6171468282585066
Test accuracy: 0.10062111914157867
Persist model completely in 'models/fcc_v1'.
(venv) potato:approach2-1 vngu$
```

mean squared error!



# 3 Different Approaches

**Approach-1**

[X]

**Approach-2**

[X]

**Approach-3**

[X]

# Given Dataset & Interpretation

dev

id	original	edit	grades	meanGrade
1723	Thousands of gay and bisexual <men/> convicted of long-abolished sexual offences are posthumously pardoned	swans	22100	1.0
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[“Special **prosecutor** appointed to Trump Russia”] + [“Special **chef** appointed to Trump Russia”] → [“0.8”]

# Preprocessing Dataset

- I. **Unfold dataset** (accordingly to interpretation)
  - A. [ “Special **prosecutor** appointed to Trump Russia” ] + [ “Special **chef** appointed to Trump Russia” ] → [ “0.8” ]
- II. **Stemming/Lemmatize & Tokenize according to word frequency**
  - A. [ 12, 833, 401, 6, 7, 9 ] + [ 12, 1050, 401, 6, 7, 9 ] → [ “0.8” ]
- III. **Padding sequences**
  - A. [ 0,0,0, 12, 833, 401, 6, 7, 9 ] + [ 0,0,0, 12, 1050, 401, 6, 7, 9 ] → [ “0.8” ]
- IV. **Normalize dataset** ( and join )
  - A. [ 0,0,0, 12/n, 833/n, 401/n, 6/n, 7/n, 9/n, 0,0,0, 12/n, 1050/n, 401/n, 6/n, 7/n, 9/n ] → [ “0.8” ]

# Network - Architecture

```
17 Parameters
1  embed_dim = 25 * 2
2  batch_size = 32
3  epochs = 100
4  num_classes = 31
5
6  # Model
7  model = Sequential()
8  model.add(Dense(256, activation='relu', input_dim=embed_dim))
9  model.add(Dropout(0.5))
10 model.add(Dense(512, activation='relu', input_dim=embed_dim))
11 model.add(Dropout(0.5))
12 model.add(Dense(num_classes, activation='softmax'))
13 model.compile(loss='categorical_crossentropy', optimizer='adam', metrics=['accuracy'])
14 print(model.metrics_names)
15
```

# Results

```
Epoch 96/100
8690/8690 [=====] - 1s 61us/step - loss: 2.1714 - accuracy: 0.2453 - val_loss: 2.6208 - val_accuracy: 0.1108
Epoch 97/100
8690/8690 [=====] - 1s 59us/step - loss: 2.1754 - accuracy: 0.2427 - val_loss: 2.6204 - val_accuracy: 0.1128
Epoch 98/100
8690/8690 [=====] - 1s 59us/step - loss: 2.1750 - accuracy: 0.2406 - val_loss: 2.6238 - val_accuracy: 0.1180
Epoch 99/100
8690/8690 [=====] - 1s 67us/step - loss: 2.1713 - accuracy: 0.2379 - val_loss: 2.6211 - val_accuracy: 0.1128
Epoch 100/100
8690/8690 [=====] - 1s 59us/step - loss: 2.1772 - accuracy: 0.2381 - val_loss: 2.6163 - val_accuracy: 0.1149
2415/2415 [=====] - 0s 13us/step
Test loss: 2.6561957538991736
Test accuracy: 0.11594203114509583
Persist model completely in 'models/fcc_v1'.
(venv) potato:approach3 vngu$
```

mean squared error!

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# Evaluation & Conclusion

# Conclusion

- Approach 1 ( and 1-1)
  - ...
- Approach 2 ( and 2-1)
  - ...
- Approach 3 most promising
  - Training-Dataset
    - 0.23 accuracy on train
    - 1.4754 on loss
  - Test- Dataset
    - 0.11 accuracy test
    - 1.6298 on loss
  - (vs. expected value of 3%)

- Current Rank 1: 0.512 train loss
- FCC works just as well as LSTM (but a lot faster)

Sub-Task 1 Results (ignore when viewing results)						
#	User	Entries	Date of Last Entry	Team Name	RMSE ▲	RM
1	HonoMi	39	02/20/20		0.51276 (1)	0.5
2	alonzorz	27	01/09/20	Amobee	0.51568 (2)	0.5
3	vgtomahawk	72	01/04/20	Hasyatarangini	0.51622 (3)	0.5
4	BramVanroy	32	12/31/19		0.51800 (4)	0.5

# Conclusion

- Approach 1 ( and 1-1)
  - ...
- Approach 2 ( and 2-1)
  - ...
- Approach 3 most promising
  - Training-Dataset
    - 0.23 accuracy on train
    - 1.4754 on loss
  - Test- Dataset
    - 0.11 accuracy test
    - 1.6298 on loss
  - (vs. expected value of 3%)
- Current Rank 1: 0.512 train loss
- FCC works just as well as LSTM (but a lot faster)

Questions? :)