Chatbot Intent Classification Modelling

We train a model that can classify speech intent for travel chatbot applications.

!pip install flat-table

Collecting flat-table
 Downloading https://files.pythonhosted.org/packages/fe/e8/5054b78b999d81acfec3606a2fcc
 Requirement already satisfied: pandas in /usr/local/lib/python3.6/dist-packages (from fl
 Requirement already satisfied: python-dateutil>=2.6.1 in /usr/local/lib/python3.6/dist-packages (f
 Requirement already satisfied: pytz>=2017.2 in /usr/local/lib/python3.6/dist-packages (f
 Requirement already satisfied: numpy>=1.13.3 in /usr/local/lib/python3.6/dist-packages (f
 Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.6/dist-packages (from
 Installing collected packages: flat-table
 Successfully installed flat-table-1.1.0

Import Relevant Libraries

travel_df = pd.read_json(path/'travel_intents.json')
chat_df = pd.read_json(path/'human-chat.json')

Process travel_df

travel_df.head()



intents

- **0** {'tag': 'greeting', 'patterns': ['Hi there', '...
- 1 {'tag': 'goodbye', 'patterns': ['Bye', 'See yo...
- 2 {'tag': 'thanks', 'patterns': ['Thanks', 'Than...
- 3 {'tag': 'options', 'patterns': ['How can help ...
- 4 {'tag': 'SearchFlight', 'patterns': ['one way ...

travel_df_flat = flat_table.normalize(travel_df, expand_lists=False)

travel_df_flat

→		index	intents.context	intents.responses	intents.patterns	intents.tag
	0	0	0	[Hello, how can I help you?, Good to see you a	[Hi there, hi there, Hello?, hi can you help m	greeting
	1	1	0	[See you!, Have a nice day, Bye! Come back aga	[Bye, See you later, bye bye, later, I will se	goodbye
	2	2		[Happy to help!, Any time!, My pleasure, No pr	[Thanks, Thank you, thanks, many thanks, thank	thanks

travel_df_flat.columns = travel_df_flat.columns.str.replace('.', '_')

travel_df_flat.drop(columns='index', inplace=True)

travel_df_flat

\Rightarrow	intents_cont	ext	intents_responses	intents_patterns	intents_tag	
	0	[]	[Hello, how can I help you?, Good to see you a	[Hi there, hi there, Hello?, hi can you help m	greeting	
	1	[]	[See you!, Have a nice day, Bye! Come back aga	[Bye, See you later, bye bye, later, I will se	goodbye	
	2	[]	[Happy to help!, Any time!, My pleasure, No pr	[Thanks, Thank you, thanks, many thanks, thank	thanks	
	3	[]	[I can provide you with flights and hotel info	[How can help me?, What you can do?, What help	options	
<pre>travel_df_flat.columns = [</pre>						

Reformat the dataframe to have context, response, text, and label as their columns.

```
lst_col = 'text'
r = pd.DataFrame({
    col:np.repeat(travel_df_flat[col].values, travel_df_flat[lst_col].str.len())
    for col in travel_df_flat.columns.drop(lst_col)}
    ).assign(**{lst_col:np.concatenate(travel_df_flat[lst_col].values)})[travel_df_flat.columns.drop(lst_col)]
```

r.head()

→		context	response	text	label
	0		[Hello, how can I help you?, Good to see you a	Hi there	greeting
	1		[Hello, how can I help you?, Good to see you a	hi there	greeting
	2		[Hello, how can I help you?, Good to see you a	Hello?	greeting
	3		[Hello, how can I help you?, Good to see you a	hi can you help me?	greeting
	4		[Hello, how can I help you?, Good to see you a	hey again	greeting

Output some general stats.

```
r['label'].value_counts()
```

```
SearchFlight 62
SearchHotelPrice 39
```

```
greeting 33
thanks 26
goodbye 24
options 14
Name: label, dtype: int64
```

```
travel_intents = r[['label','text']]
```

```
r.to_csv(path/'travel_chat.csv', index=False)
```

Train-test split the travel_intent dataframe.

```
travel_intents_copy = travel_intents.copy()
travel_train_df = travel_intents_copy.sample(frac=0.7, random_state=42)
travel_val_df = travel_intents_copy.drop(travel_train_df.index)
```

Process chat_df

This is the corpus for general telegram chat logs.

chat_df.head()

→	id		sender	text	evaluation_score	sender_class
	0	1	participant1	Thats good for you, i'm not very into new tech	NaN	Human
	1	3	participant1	I am go to gym and live on donations	NaN	Human
	2	5	participant1	So vegan i have dogs maybe i should told th	NaN	Human
	3	7	participant1	Dogs or vegan in office?	NaN	Human

```
chat_df.describe()
```



count 18867.000000 0.0 12.479408 NaN mean std 14.337071 NaN min 0.000000 NaN 25% 3.000000 NaN 50% 9.000000 NaN 75% 17.000000 NaN 175.000000 NaN max

id evaluation_score

chat_train_df.describe()

□ label text

count	13207	13207
unique	1	9819
top	Human	Hello
freq	13207	158

```
lm_train = pd.concat([chat_train_df, travel_train_df], ignore_index=True)
lm_val = pd.concat([chat_val_df, travel_val_df], ignore_index=True)

# Shuffle the rows
lm_train = lm_train.sample(frac=1).reset_index(drop=True)
lm_val = lm_val.sample(frac=1).reset_index(drop=True)

lm train.describe()
```



	label	text
count	13346	13346
unique	7	9939
top	Human	Hello
freq	13207	159

Fine tune Language Model

Define data bunches

```
# Langauge modeling data bunch
data_lm = TextLMDataBunch.from_df(path, train_df = lm_train, valid_df = lm_val, bs =32)
→
data lm
   TextLMDataBunch;
     Train: LabelList (13346 items)
     x: LMTextList
     xxbos thanks,xxbos i could do it for you for free ! i enjoy xxunk .,xxbos i 'm a
     housekeeper,xxbos i think i might started xxunk weight when my wife left me ...,xxbos
     xxmaj where do you live ? xxmaj close to mountains ?
     y: LMLabelList
     , , , ,
     Path: /content/gdrive/My Drive/Colab
     Notebooks/1000ML/Unit6_Neural_Networks/Projects/Travel_Chatbot;
     Valid: LabelList (5719 items)
     x: LMTextList
     xxbos xxmaj and i like to eat you know .,xxbos i really do need some love,xxbos xxmaj
     nice . xxmaj why fiction ?,xxbos i am good , and you ? xxmaj just spending time with my
     family , as i love to,xxbos xxmaj hi , how are you ? xxmaj no , i can make any number
     of them for myself any time i want )
     y: LMLabelList
     ,,,,
     Path: /content/gdrive/My Drive/Colab
     Notebooks/1000ML/Unit6 Neural Networks/Projects/Travel Chatbot;
     Test: None
data_lm.show_batch()
```



idx text

texas where are you from ? xxbos xxmaj yes xxbos i should go , bye xxbos what is xxunk xxunk ? xxbos xxmaj where are you from xxbos xxmaj what is your name ? xxbos i m feeling fine xxbos sorry to hear that xxbos xxup okay xxbos xxmaj okay . xxmaj do you want to buy a car ? xxbos my current goal is to run a k xxbos

- lot of instruments as well xxbos xxmaj well shit xxbos xxmaj makes sense xxbos xxmaj what is your favorite band ? xxbos yeah , what are you going to do this weekend ? xxbos xxmaj hi ! xxmaj what a nice day ! xxmaj how are you ? xxbos xxmaj what else do you like ? xxbos xxmaj already no xxbos xxmaj hi , i 'm fine . xxmaj you
- training, more training, some training. xxmaj usual stuff. xxbos where are u ? xxbos xxmaj that 's a job xxbos xxmaj hello! xxbos i really like to travel. xxbos xxmaj great. xxmaj do you know many languages? xxbos playing piano?:) xxbos xxmaj no, i do n't. xxmaj do you have a dog? xxbos i 'd say xxbos

learn = language_model_learner(data_lm, AWD_LSTM, drop_mult=0.5, pretrained=True)

Downloading https://s3.amazonaws.com/fast-ai-modelzoo/wt103-fwd

learn.lr find()



66.67% [2/3 00:11<00:05]

epoch	train_loss	valid_loss	accuracy	time
0	6.535960	#na#	00:05	
1	5.313480	#na#	00:05	

20.00% [9/45 00:01<00:04 7.7893]

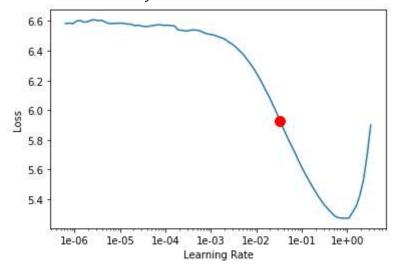
LR Finder is complete, type {learner_name}.recorder.plot() to see the graph.

Discrimitive fine-tuning

learn.recorder.plot(suggestion=True)



Min numerical gradient: 3.31E-02 Min loss divided by 10: 9.12E-02



learn.fit_one_cycle(4, 9.12E-02,callbacks=[SaveModelCallback(learn, name="best_lm")], moms=(

→	epoch	train_loss	valid_loss	accuracy	time
	0	3.856387	3.186715	0.361518	00:06
	1	3.386144	3.035699	0.384732	00:06
	2	3.053555	2.888120	0.399487	00:06
	3	2.758148	2.838932	0.407433	00:06

Better model found at epoch 0 with valid_loss value: 3.1867151260375977. Better model found at epoch 1 with valid_loss value: 3.0356991291046143. Better model found at epoch 2 with valid_loss value: 2.888120174407959. Better model found at epoch 3 with valid_loss value: 2.8389320373535156.

learn.save('fit_head')

Train lower layers

learn.unfreeze()

learn.lr_find()

learn.recorder.plot(suggestion=True)



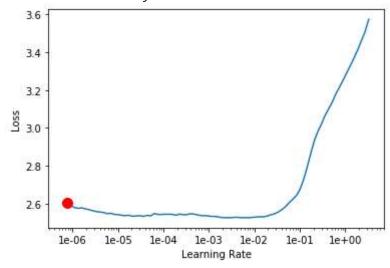
66.67% [2/3 00:12<00:06]

epoch	train_loss	valid_loss	accuracy	time
0	2.541485	#na#	00:06	
1	3.332014	#na#	00:06	

20.00% [9/45 00:01<00:04 4.1721]

LR Finder is complete, type {learner_name}.recorder.plot() to see the graph.

Min numerical gradient: 7.59E-07 Min loss divided by 10: 7.59E-04



learn.fit_one_cycle(1,2E-06,callbacks=[SaveModelCallback(learn, name="best_lm")], moms=(0.8,



epoch	train_loss	valid_loss	accuracy	time
0	2.497512	2.838806	0.407455	00:07

Better model found at epoch 0 with valid_loss value: 2.838805675506592.

learn.load('best_lm')



```
(module): LSTM(400, 1152, batch_first=True)
           )
           (1): WeightDropout(
             (module): LSTM(1152, 1152, batch_first=True)
           (2): WeightDropout(
             (module): LSTM(1152, 400, batch_first=True)
         )
         (input_dp): RNNDropout()
         (hidden dps): ModuleList(
           (0): RNNDropout()
           (1): RNNDropout()
           (2): RNNDropout()
         )
       (1): LinearDecoder(
         (decoder): Linear(in features=400, out features=2032, bias=True)
         (output dp): RNNDropout()
       )
     ), opt_func=functools.partial(<class 'torch.optim.adam.Adam'>, betas=(0.9, 0.99)),
     loss_func=FlattenedLoss of CrossEntropyLoss(), metrics=[<function accuracy at
     0x7ff9405d60d0>], true wd=True, bn wd=True, wd=0.01, train bn=True,
     path=PosixPath('/content/gdrive/My Drive/Colab
     Notebooks/1000ML/Unit6 Neural Networks/Projects/Travel Chatbot'), model dir='models',
     callback_fns=[functools.partial(<class 'fastai.basic_train.Recorder'>, add_time=True,
     silent=False)], callbacks=[...], layer_groups=[Sequential(
       (0): WeightDropout(
         (module): LSTM(400, 1152, batch_first=True)
       (1): RNNDropout()
     ), Sequential(
       (0): WeightDropout(
         (module): LSTM(1152, 1152, batch_first=True)
       (1): RNNDropout()
     ), Sequential(
       (0): WeightDropout(
         (module): LSTM(1152, 400, batch_first=True)
       (1): RNNDropout()
# Save the fine-tuned encoder to be used later for fine-tuning the classifier
learn.save_encoder('ft_enc')
learn.predict("hi", n_words=20)
\overrightarrow{
ightarrow} 'hi i 'm a vegan ! xxbos What is Delta ? xxbos navy blue now , and u mom'
```

Train Text Classifier

Classififer model data using just the intents file data_clas = TextClasDataBunch.from_df(path, train_df = travel_train_df, valid_df = travel_va vocab=data_lm.train_ds.vocab, bs=8)



data_clas

TextClasDataBunch;

Train: LabelList (139 items)

x: TextList

xxbos xxmaj awesome , thanks,xxbos xxmaj one way to xxmaj xxunk from xxmaj hong xxmaj kong, xxbos xxmaj hello, xxbos xxmaj roundtrip flight between xxmaj san xxmaj jose and xxmaj new xxmaj york xxmaj city leaving on xxmaj feb 2 and returning on xxmaj feb 5 ?,xxbos xxmaj looking for roundtrip price from xxmaj paris to xxmaj xxunk

y: CategoryList

thanks, SearchFlight, greeting, SearchFlight, SearchFlight

Path: /content/gdrive/My Drive/Colab

Notebooks/1000ML/Unit6 Neural Networks/Projects/Travel Chatbot;

Valid: LabelList (59 items)

x: TextList

xxbos hi there,xxbos hello ? anyone ?,xxbos hi again,xxbos hi there , can you help me ?,xxbos hello , good day

y: CategoryList

greeting, greeting, greeting, greeting

Path: /content/gdrive/My Drive/Colab

Notebooks/1000ML/Unit6 Neural Networks/Projects/Travel Chatbot;

Test: None

data_clas.show_batch()



text target xxbos xxmaj price for roundtrip from xxmaj toronto to xxmaj new xxmaj york xxmaj city if i SearchFlight leave on xxmaj feb 2 and coming back on xxmaj feb 5? xxbos xxmaj price for trip from xxmaj san xxmaj jose to xxmaj new xxmaj york xxmaj city SearchFlight leaving on xxmaj feb 2 and returning on xxmaj feb 5? xxbos xxmaj roundtrip flight between xxmaj san xxmaj jose and xxmaj new xxmaj york SearchFlight xxmaj city leaving on xxmaj feb 2 and returning on xxmaj feb 5? xxbos xxmai flight from xxmai san xxmai iose to xxmai new xxmai vork xxmai city leaving clas learn = text classifier learner(data clas, AWD LSTM, drop mult=0.5) clas learn.load encoder('ft enc')

RNNLearner(data=TextClasDataBunch;

```
Train: LabelList (139 items)
x: TextList
xxbos xxmaj awesome , thanks,xxbos xxmaj one way to xxmaj xxunk from xxmaj hong xxmaj
kong, xxbos xxmaj hello, xxbos xxmaj roundtrip flight between xxmaj san xxmaj jose and
xxmaj new xxmaj york xxmaj city leaving on xxmaj feb 2 and returning on xxmaj feb 5
?,xxbos xxmaj looking for roundtrip price from xxmaj paris to xxmaj xxunk
y: CategoryList
thanks, SearchFlight, greeting, SearchFlight, SearchFlight
Path: /content/gdrive/My Drive/Colab
Notebooks/1000ML/Unit6 Neural Networks/Projects/Travel Chatbot;
Valid: LabelList (59 items)
x: TextList
xxbos hi there,xxbos hello ? anyone ?,xxbos hi again,xxbos hi there , can you help me
?,xxbos hello , good day
y: CategoryList
greeting, greeting, greeting, greeting
Path: /content/gdrive/My Drive/Colab
Notebooks/1000ML/Unit6_Neural_Networks/Projects/Travel_Chatbot;
Test: None, model=SequentialRNN(
  (0): MultiBatchEncoder(
    (module): AWD LSTM(
      (encoder): Embedding(2032, 400, padding idx=1)
      (encoder dp): EmbeddingDropout(
        (emb): Embedding(2032, 400, padding idx=1)
      )
      (rnns): ModuleList(
        (0): WeightDropout(
          (module): LSTM(400, 1152, batch_first=True)
        (1): WeightDropout(
          (module): LSTM(1152, 1152, batch first=True)
        (2): WeightDropout(
          (module): LSTM(1152, 400, batch_first=True)
        )
      (input dp): RNNDropout()
      (hidden_dps): ModuleList(
        (0): RNNDropout()
        (1): RNNDropout()
        (2): RNNDropout()
      )
    )
  (1): PoolingLinearClassifier(
    (layers): Sequential(
      (0): BatchNorm1d(1200, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
      (1): Dropout(p=0.2, inplace=False)
      (2): Linear(in features=1200, out features=50, bias=True)
      (3): ReLU(inplace=True)
      (4): BatchNorm1d(50, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
      (5) · Dronout(n=0 1 innlace=False)
```

clas_learn.lr_find()
clas_learn.recorder.plot(suggestion=True)



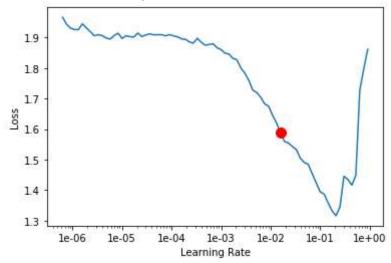
83.33% [5/6 00:02<00:00]

time	accuracy	valid_loss	train_loss	epoch
	00:00	#na#	1.931725	0
	00:00	#na#	1.909148	1
	00:00	#na#	1.860383	2
	00:00	#na#	1.554657	3
	00:00	#na#	1.448750	4

29.41% [5/17 00:00<00:00 2.7734]

LR Finder is complete, type {learner_name}.recorder.plot() to see the graph.

Min numerical gradient: 1.58E-02 Min loss divided by 10: 2.09E-02



best_clf_lr = clas_learn.recorder.min_grad_lr
best_clf_lr

0.01584893192461114

clas_learn.fit_one_cycle(4, 1.5E-2)

→	epoch	train_loss	valid_loss	accuracy	time
	0	1.372939	0.924876	0.830508	00:00
	1	1.102432	0.622292	0.881356	00:00
	2	0.922053	0.430753	0.932203	00:00
	3	0.828463	0.369007	0.915254	00:00

Train the last 2 layers

clas_learn.freeze_to(-2)

clas_learn.lr_find()

clas_learn.recorder.plot(suggestion=True)



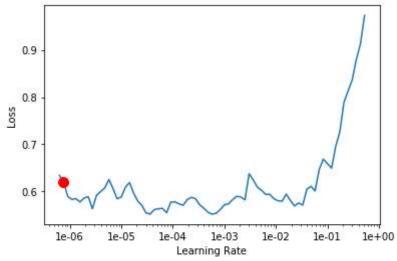
83.33% [5/6 00:02<00:00]

epoch	train_loss	<pre>valid_loss</pre>	accuracy	time
0	0.586332	#na#	00:00	
1	0.561878	#na#	00:00	
2	0.571751	#na#	00:00	
3	0.569516	#na#	00:00	
4	0.973189	#na#	00:00	

23.53% [4/17 00:00<00:00 1.9005]

LR Finder is complete, type {learner_name}.recorder.plot() to see the graph.

Min numerical gradient: 7.59E-07 Min loss divided by 10: 5.75E-05



clas_learn.fit_one_cycle(1, 1.e-6)

Train the rest

clas_learn.unfreeze()
clas_learn.lr_find()

clas_learn.recorder.plot(suggestion=True)

₹

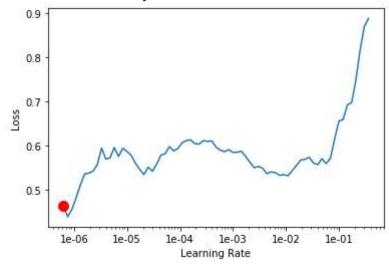
83.33% [5/6 00:03<00:00]

epoch	train_loss	valid_loss	accuracy	time
0	0.537826	#na#	00:00	
1	0.578277	#na#	00:00	
2	0.584453	#na#	00:00	
3	0.568885	#na#	00:00	
4	1.025922	#na#	00:00	

11.76% [2/17 00:00<00:01 1.2329]

LR Finder is complete, type {learner_name}.recorder.plot() to see the graph.

Min numerical gradient: 6.31E-07 Min loss divided by 10: 7.59E-08



best_clf_lr = clas_learn.recorder.min_grad_lr
best_clf_lr

6.309573444801933e-07

clas_learn.fit_one_cycle(1, 1e-7)

epoch train_loss valid_loss accuracy time

0 0.544133 0.369690 0.932203 00:00

Export classification model.

clas_learn.save('travel-chat-clas-model')

Do some tests.

```
clas_learn.predict("going on a trip to london")
→ (Category SearchFlight,
      tensor(0),
      tensor([0.7810, 0.0092, 0.0532, 0.1277, 0.0191, 0.0098]))
clas learn.predict("Need a hotel in london")
→ (Category SearchHotelPrice,
      tensor(1),
      tensor([0.0722, 0.7887, 0.0729, 0.0188, 0.0298, 0.0175]))
clas learn.predict("What can you do for me")
→ (Category options,
      tensor(4),
      tensor([0.0143, 0.0015, 0.0456, 0.0803, 0.7493, 0.1090]))
clas learn.predict("hello")
→ (Category greeting,
      tensor(3),
      tensor([1.8479e-03, 1.8565e-03, 1.5479e-03, 9.7622e-01, 1.7667e-02, 8.6011e-04]))
clas_learn.predict("thanks alot, much appreicated")
→ (Category thanks,
      tensor(5),
      tensor([1.4571e-04, 6.2914e-04, 1.6595e-04, 3.7264e-05, 5.6626e-04, 9.9846e-01]))
Look at the confusion matrix to see the breakdown of results.
from fastai.vision import ClassificationInterpretation
interp = ClassificationInterpretation.from_learner(clas_learn)
\rightarrow
interp.plot_confusion_matrix(figsize=(12,12), dpi=60)
```



