

RNN_model_imdbreviews

May 9, 2021

Recurrent Neural Network (RNN)

Import Libraries

```
[1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
```

```
import tensorflow as tf
from scipy.spatial.distance import cdist
```

```
[2]: from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense, GRU, Embedding
from tensorflow.keras.optimizers import Adam
from tensorflow.keras.preprocessing.text import Tokenizer
from tensorflow.keras.preprocessing.sequence import pad_sequences
```

```
[3]: import tensorflow_hub as hub
import tensorflow_datasets as tfds
```

Access and split the dataset

```
[4]: train_data, validation_data, test_data = tfds.load(
    name = "imdb_reviews",
    split=('train[:60%]', 'train[60%:]', 'test'),
    as_supervised=True
)
```

Downloading and preparing dataset 80.23 MiB (download: 80.23 MiB, generated:

Unknown size, total: 80.23 MiB) to

/home/jayanthikishore/tensorflow_datasets/imdb_reviews/plain_text/1.0.0...

HBox(children=(FloatProgress(value=1.0, bar_style='info', description='Dl Completed...', max=1

HBox(children=(FloatProgress(value=1.0, bar_style='info', description='Dl Size...', max=1.0, s

```
HBox(children=(FloatProgress(value=0.0, description='Generating splits...', max=3.0, style=Pro
```

```
HBox(children=(FloatProgress(value=0.0, description='Generating train examples...', max=25000.0
```

```
HBox(children=(FloatProgress(value=0.0, description='Shuffling imdb_reviews-train.tfrecord...'
```

```
HBox(children=(FloatProgress(value=0.0, description='Generating test examples...', max=25000.0
```

```
HBox(children=(FloatProgress(value=0.0, description='Shuffling imdb_reviews-test.tfrecord...'
```

```
HBox(children=(FloatProgress(value=0.0, description='Generating unsupervised examples...', max
```

```
HBox(children=(FloatProgress(value=0.0, description='Shuffling imdb_reviews-unsupervised.tfrec
```

Dataset imdb_reviews downloaded and prepared to

/home/jayanthikishore/tensorflow_datasets/imdb_reviews/plain_text/1.0.0.

Subsequent calls will reuse this data.

```
[5]: train_example, train_labels = next(iter(train_data.batch(15)))
      train_example
```

```
[5]: <tf.Tensor: shape=(15,), dtype=string, numpy=
array([b"This was an absolutely terrible movie. Don't be lured in by Christopher
Walken or Michael Ironside. Both are great actors, but this must simply be their
worst role in history. Even their great acting could not redeem this movie's
ridiculous storyline. This movie is an early nineties US propaganda piece. The
most pathetic scenes were those when the Columbian rebels were making their
cases for revolutions. Maria Conchita Alonso appeared phony, and her pseudo-love
affair with Walken was nothing but a pathetic emotional plug in a movie that was
devoid of any real meaning. I am disappointed that there are movies like this,
ruining actor's like Christopher Walken's good name. I could barely sit through
it."],
```

```
      b'I have been known to fall asleep during films, but this is usually due
to a combination of things including, really tired, being warm and comfortable
on the sette and having just eaten a lot. However on this occasion I fell asleep
because the film was rubbish. The plot development was constant. Constantly slow
and boring. Things seemed to happen, but with no explanation of what was causing
them or why. I admit, I may have missed part of the film, but i watched the
majority of it and everything just seemed to happen of its own accord without
any real concern for anything else. I cant recommend this film at all.',
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```
      b'Mann photographs the Alberta Rocky Mountains in a superb fashion, and
Jimmy Stewart and Walter Brennan give enjoyable performances as they always seem
to do. <br /><br />But come on Hollywood - a Mountie telling the people of
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Dawson City, Yukon to elect themselves a marshal (yes a marshal!) and to enforce the law themselves, then gunfighters battling it out on the streets for control of the town?

Nothing even remotely resembling that happened on the Canadian side of the border during the Klondike gold rush. Mr. Mann and company appear to have mistaken Dawson City for Deadwood, the Canadian North for the American Wild West.

Canadian viewers be prepared for a Reefer Madness type of enjoyable howl with this ludicrous plot, or, to shake your head in disgust.',

b'This is the kind of film for a snowy Sunday afternoon when the rest of the world can go ahead with its own business as you descend into a big arm-chair and mellow for a couple of hours. Wonderful performances from Cher and Nicolas Cage (as always) gently row the plot along. There are no rapids to cross, no dangerous waters, just a warm and witty paddle through New York life at its best. A family film in every sense and one that deserves the praise it received.',

b'As others have mentioned, all the women that go nude in this film are mostly absolutely gorgeous. The plot very ably shows the hypocrisy of the female libido. When men are around they want to be pursued, but when no "men" are around, they become the pursuers of a 14 year old boy. And the boy becomes a man really fast (we should all be so lucky at this age!). He then gets up the courage to pursue his true love.',

b"This is a film which should be seen by anybody interested in, effected by, or suffering from an eating disorder. It is an amazingly accurate and sensitive portrayal of bulimia in a teenage girl, its causes and its symptoms. The girl is played by one of the most brilliant young actresses working in cinema today, Alison Lohman, who was later so spectacular in 'Where the Truth Lies'. I would recommend that this film be shown in all schools, as you will never see a better on this subject. Alison Lohman is absolutely outstanding, and one marvels at her ability to convey the anguish of a girl suffering from this compulsive disorder. If barometers tell us the air pressure, Alison Lohman tells us the emotional pressure with the same degree of accuracy. Her emotional range is so precise, each scene could be measured microscopically for its gradations of trauma, on a scale of rising hysteria and desperation which reaches unbearable intensity. Mare Winningham is the perfect choice to play her mother, and does so with immense sympathy and a range of emotions just as finely tuned as Lohman's. Together, they make a pair of sensitive emotional oscillators vibrating in resonance with one another. This film is really an astonishing achievement, and director Katt Shea should be proud of it. The only reason for not seeing it is if you are not interested in people. But even if you like nature films best, this is after all animal behaviour at the sharp edge. Bulimia is an extreme version of how a tormented soul can destroy her own body in a frenzy of despair. And if we don't sympathise with people suffering from the depths of despair, then we are dead inside.",

b'Okay, you have:

Penelope Keith as Miss Herringbone-Tweed, B.B.E. (Backbone of England.) She\'s killed off in the first scene - that\'s right, folks; this show has no backbone!

Peter O\'Toole as Ol\' Colonel Cricket from The First War and now the emblazered Lord of the Manor.

Joanna Lumley as the ensweatered Lady of the Manor, 20 years younger than the colonel and 20 years past her own prime but still glamorous (Brit spelling, not mine) enough to have a toy-boy on the side. It\'s alright, they have Col. Cricket\'s full knowledge and consent (they guy even comes \'round for Christmas!) Still, she\'s considerate of the colonel enough to have said toy-boy her own age (what a gal!)

David McCallum as said toy-boy, equally as pointlessly glamorous as his squeeze. Pilcher couldn\'t come up with any cover for him within the story, so she gave him a hush-hush job at the Circus.

and finally:

Susan Hampshire as Miss Polonia Teacups, Venerable Headmistress of the Venerable Girls\' Boarding-School, serving tea in her office with a dash of deep, poignant advice for life in the outside world just before graduation. Her best bit of advice: "I\'ve only been to Nancherrow (the local Stately Home of England) once. I thought it was very beautiful but, somehow, not part of the real world." Well, we can\'t say they didn\'t warn us.

Ah, Susan - time was, your character would have been running the whole show. They don\'t write \'em like that any more. Our loss, not yours.

So - with a cast and setting like this, you have the re-makings of "Brideshead Revisited," right?

Wrong! They took these 1-dimensional supporting roles because they paid so well. After all, acting is one of the oldest temp-jobs there is (YOU name another!)

First warning sign: lots and lots of backlighting. They get around it by shooting outdoors - "hey, it\'s just the sunlight!"

Second warning sign: Leading Lady cries a lot. When not crying, her eyes are moist. That\'s the law of romance novels: Leading Lady is "dewy-eyed."

Henceforth, Leading Lady shall be known as L.L.

Third warning sign: L.L. actually has stars in her eyes when she\'s in love. Still, I\'ll give Emily Mortimer an award just for having to act with that spotlight in her eyes (I wonder . did they use contacts?)

And lastly, fourth warning sign: no on-screen female character is "Mrs." She\'s either "Miss" or "Lady."

When all was said and done, I still couldn\'t tell you who was pursuing whom and why. I couldn\'t even tell you what was said and done.

To sum up: they all live through World War II without anything happening to them at all.

OK, at the end, L.L. finds she\'s lost her parents to the Japanese prison camps and baby sis comes home catatonic. Meanwhile (there\'s always a "meanwhile,") some young guy L.L. had a crush on (when, I don\'t know) comes home from some wartime tough spot and is found living on the street by Lady of the Manor (must be some street if SHE\'s going to find him there.) Both war casualties are whisked away to recover at Nancherrow (SOMEBODY has to be "whisked away" SOMEWHERE in these romance stories!)

Great drama.',

b'The film is based on a genuine 1950s novel.

Journalist Colin McInnes wrote a set of three "London novels": "Absolute Beginners", "City of Spades" and "Mr Love and Justice". I have read all three. The first two are excellent. The last, perhaps an experiment that did not come off. But McInnes\'s work is highly acclaimed; and rightly so. This musical is the novelist\'s ultimate nightmare - to see the fruits of one\'s mind being turned into a glitzy, badly-acted, soporific one-dimensional apology of a film that says it captures the spirit of 1950s London, and does nothing of the sort.

/>Thank goodness Colin McInnes wasn't alive to witness it.',

b'I really love the sexy action and sci-fi films of the sixties and its because of the actress's that appeared in them. They found the sexiest women to be in these films and it didn't matter if they could act (Remember "Candy"?). The reason I was disappointed by this film was because it wasn't nostalgic enough. The story here has a European sci-fi film called "Dragonfly" being made and the director is fired. So the producers decide to let a young aspiring filmmaker (Jeremy Davies) to complete the picture. They're is one real beautiful woman in the film who plays Dragonfly but she's barely in it. Film is written and directed by Roman Coppola who uses some of his fathers exploits from his early days and puts it into the script. I wish the film could have been an homage to those early films. They could have lots of cameos by actors who appeared in them. There is one actor in this film who was popular from the sixties and its John Phillip Law (Barbarella). Gerard Depardieu, Giancarlo Giannini and Dean Stockwell appear as well. I guess I'm going to have to continue waiting for a director to make a good homage to the films of the sixties. If any are reading this, "Make it as sexy as you can"! I'll be waiting!',

b'Sure, this one isn't really a blockbuster, nor does it target such a position. "Dieter" is the first name of a quite popular German musician, who is either loved or hated for his kind of acting and thats exactly what this movie is about. It is based on the autobiography "Dieter Bohlen" wrote a few years ago but isn't meant to be accurate on that. The movie is filled with some sexual offensive content (at least for American standard) which is either amusing (not for the other "actors" of course) or dumb - it depends on your individual kind of humor or on you being a "Bohlen"-Fan or not. Technically speaking there isn't much to criticize. Speaking of me I find this movie to be an OK-movie.',

b'During a sleepless night, I was switching through the channels & found this embarrassment of a movie. What were they thinking?

If this is life after "Remote Control" for Kari (Wuhrer) Salin, no wonder she's gone nowhere.

And why did David Keith take this role? It's pathetic!

Anyway, I turned on the movie near the end, so I didn't get much of the plot. But this must've been the best part. This nerdy college kid brings home this dominatrix-ish girl...this scene is straight out of the comic books -- or the cheap porn movies. She calls the mother anal retentive and kisses the father "Oh, I didn't expect tongue!" Great lines!

After this, I had to see how it ended..

Well, of course, this bitch from hell has a helluva past, so the SWAT team is upstairs. And yes...they surround her! And YES YES! The kid blows her brains out!!!! AHAHAHAHAHA!!

This is must-see TV.

',

b'Cute film about three lively sisters from Switzerland (often seen running about in matching outfits) who want to get their parents back together (seems mom is still carrying the torch for dad) - so they sail off to New York to stop the dad from marrying a blonde gold-digger he calls "Precious". Dad hasn't seen his daughters in ten years, they (oddly enough) don't seem to mind and think he's wonderful, and meanwhile Precious seems to lead a life mainly run by her overbearing mother (Alice Brady), a woman who just wants to see to it

her daughter marries a rich man. The sisters get the idea of pushing Precious into the path of a drunken Hungarian count, tricking the two gold-digging women into thinking he is one of the richest men in Europe. But a case of mistaken identity makes the girls think the count is good-looking Ray Milland, who goes along with the scheme 'cause he has a crush on sister Kay.

This film is enjoyable, light fare. Barbara Read as Kay comes across as sweet and pretty, Ray Milland looks oh so young and handsome here (though, unfortunately, is given little to do), Alice Brady is quite good as the scheming mother - but it is Deanna Durbin, a real charmer and cute as a button playing youngest sister Penny, who pretty much steals the show. With absolutely beautiful vocals, she sings several songs throughout the film, though I actually would have liked to have seen them feature her even more in this. The plot in this film is a bit silly, but nevertheless, I found the film to be entertaining and fun.',

b"This 1984 version of the Dickens' classic 'A Christmas Carol,' directed by Clive Donner, stars George C. Scott as Ebenezer Scrooge. By this time around, the challenge for the filmmaker was to take such familiar material and make it seem fresh and new again; and, happily to say, with this film Donner not only met the challenge but surpassed any expectations anyone might have had for it. He tells the story with precision and an eye to detail, and extracts performances from his actors that are nothing less than superlative, especially Scott. One could argue that the definitive portrayal of Scrooge-- one of the best known characters in literary fiction, ever-- was created by Alastair Sim in the 1951 film; but I think with his performance here, Scott has now achieved that distinction. There is such a purity and honesty in his Scrooge that it becomes difficult to even consider anyone else in the role once you've seen Scott do it; simply put, he IS Scrooge. And what a tribute it is to such a gifted actor; to be able to take such a well known figure and make it so uniquely his own is quite miraculous. It is truly a joy to see an actor ply his trade so well, to be able to make a character so real, from every word he utters down to the finest expression of his face, and to make it all ring so true. It's a study in perfection.

The other members of the cast are splendid as well, but then again they have to be in order to maintain the integrity of Scott's performance; and they do. Frank Finlay is the Ghost of Jacob Marley; a notable turn, though not as memorable, perhaps, as the one by Alec Guinness (as Marley) in the film, 'Scrooge.' Angela Pleasence is a welcome visage as the Spirit of Christmas Past; Edward Woodward, grand and boisterous, and altogether convincing as the Spirit of Christmas Present; and Michael Carter, grim and menacing as the Spirit of Christmas Yet To Come.

David Warner hits just the right mark with his Bob Cratchit, bringing a sincerity to the role that measures up well to the standard of quality set by Scott's Scrooge, and Susannah York fares just as well as Mrs. Cratchit. The real gem to be found here, though, is the performance of young Anthony Walters as Tiny Tim; it's heartfelt without ever becoming maudlin, and simply one of the best interpretations-- and the most real-- ever presented on film.

The excellent supporting cast includes Roger Rees (Fred Holywell, and also the narrator of the film), Caroline Langrishe (Janet Holywell), Lucy Gutteridge (Belle), Michael Gough (Mr. Poole) and Joanne Whalley (Fan). A flawless presentation, this version of 'A Christmas

Carol' sets the standard against which all others must be gauged; no matter how many versions you may have seen, watching this one is like seeing it for the first time ever. And forever after, whenever you think of Scrooge, the image your mind will conjure up will be that of George C. Scott. A thoroughly entertaining and satisfying experience, this film demands a place in the annual schedule of the holiday festivities of every home. I rate this one 10/10.",

b'Put the blame on executive producer Wes Craven and financiers the Weinsteins for this big-budget debacle: a thrash-metal updating of "Dracula", with a condescending verbal jab at Bram Stoker (who probably wouldn't want his name on this thing anyway) and nothing much for the rest of us except slasher-styled jolts and gore. Christopher Plummer looks winded as Van Helsing in the modern-day--not just a descendant of Van Helsing but the real thing; he keeps himself going with leeches obtained from Count Dracula's corpse, which is exhumed from its coffin after being stolen from Van Helsing's vault and flown to New Orleans. This is just what New Orleans needs in the 21st Century! The film, well-produced but without a single original idea (except for multi-racial victims), is both repulsive and lazy, and after about an hour starts repeating itself. * from ****',

b'Hilarious, evocative, confusing, brilliant film. Reminds me of Bunuel's L'Age D'Or or Jodorowsky's Holy Mountain-- lots of strange characters mucking about and looking for... what is it? I laughed almost the whole way through, all the while keeping a peripheral eye on the bewildered and occasionally horrified reactions of the audience that surrounded me in the theatre. Entertaining through and through, from the beginning to the guts and poisoned entrails all the way to the end, if it was an end. I only wish i could remember every detail. It haunts me sometimes.

Honestly, though, i have only the most positive recollections of this film. As it doesn't seem to be available to take home and watch, i suppose i'll have to wait a few more years until Crispin Glover comes my way again with his Big Slide Show (and subsequent "What is it?" screening)... I saw this film in Atlanta almost directly after being involved in a rather devastating car crash, so i was slightly dazed at the time, which was perhaps a very good state of mind to watch the prophetic talking arthropods and the retards in the superhero costumes and godlike Glover in his appropriate burly-Q setting, scantily clad girlies rising out of the floor like a magnificent DADAist wet dream.

Is it a statement on Life As We Know It? Of course everyone EXPECTS art to be just that. I rather think that the truth is more evident in the absences and in the negative space. What you don't tell us is what we must deduce, but is far more valid than the lies that other people feed us day in and day out. Rather one "WHAT IS IT?" than 5000 movies like "Titanic" or "Sleepless in Seattle" (shudder, gag, groan).

Thank you, Mr. Glover (additionally a fun man to watch on screen or at his Big Slide Show-- smart, funny, quirky, and outrageously hot). Make more films, write more books, keep the nightmare alive.'],

dtype=object)>

```
[6]: tf.shape(train_example)
```

```
[6]: <tf.Tensor: shape=(1,), dtype=int32, numpy=array([15], dtype=int32)>
```

```
[7]: train_labels
```

```
[7]: <tf.Tensor: shape=(15,), dtype=int64, numpy=array([0, 0, 0, 1, 1, 1, 0, 0, 0, 0, 0, 1, 1, 0, 1])>
```

Embeddings

```
[8]: embedding = "https://tfhub.dev/google/nnlm-en-dim50/2"
```

```
[9]: hub_layer = hub.KerasLayer(embedding, input_shape = [], dtype=tf.  
    ↪string, trainable=True)  
hub_layer(train_example[:3])
```

```
[9]: <tf.Tensor: shape=(3, 50), dtype=float32, numpy=  
array([[ 0.5423195, -0.0119017,  0.06337538,  0.06862972, -0.16776837,  
        -0.10581174,  0.16865303, -0.04998824, -0.31148055,  0.07910346,  
         0.15442263,  0.01488662,  0.03930153,  0.19772711, -0.12215476,  
        -0.04120981, -0.2704109, -0.21922152,  0.26517662, -0.80739075,  
         0.25833532, -0.3100421,  0.28683215,  0.1943387, -0.29036492,  
         0.03862849, -0.7844411, -0.0479324,  0.4110299, -0.36388892,  
        -0.58034706,  0.30269456,  0.3630897, -0.15227164, -0.44391504,  
         0.19462997,  0.19528408,  0.05666234,  0.2890704, -0.28468323,  
        -0.00531206,  0.0571938, -0.3201318, -0.04418665, -0.08550783,  
        -0.55847436, -0.23336391, -0.20782952, -0.03543064, -0.17533456],  
       [ 0.56338924, -0.12339553, -0.10862679,  0.7753425, -0.07667089,  
        -0.15752277,  0.01872335, -0.08169781, -0.3521876,  0.4637341,  
        -0.08492756,  0.07166859, -0.00670817,  0.12686075, -0.19326553,  
        -0.52626437, -0.3295823,  0.14394785,  0.09043556, -0.5417555,  
         0.02468163, -0.15456742,  0.68333143,  0.09068331, -0.45327246,  
         0.23180096, -0.8615696,  0.34480393,  0.12838456, -0.58759046,  
        -0.4071231,  0.23061076,  0.48426893, -0.27128142, -0.5380916,  
         0.47016326,  0.22572741, -0.00830663,  0.2846242, -0.304985,  
         0.04400365,  0.25025874,  0.14867121,  0.40717036, -0.15422426,  
        -0.06878027, -0.40825695, -0.3149215,  0.09283665, -0.20183425],  
       [ 0.7456154,  0.21256861,  0.14400336,  0.5233862,  0.11032254,  
         0.00902788, -0.3667802, -0.08938274, -0.24165542,  0.33384594,  
        -0.11194605, -0.01460047, -0.0071645,  0.19562712,  0.00685216,  
        -0.24886718, -0.42796347,  0.18620004, -0.05241098, -0.66462487,  
         0.13449019, -0.22205497,  0.08633006,  0.43685386,  0.2972681,  
         0.36140734, -0.7196889,  0.05291241, -0.14316116, -0.1573394,  
        -0.15056328, -0.05988009, -0.08178931, -0.15569411, -0.09303783,  
        -0.18971172,  0.07620788, -0.02541647, -0.27134508, -0.3392682,  
        -0.10296468, -0.27275252, -0.34078008,  0.20083304, -0.26644835,  
         0.00655449, -0.05141488, -0.04261917, -0.45413622,  0.20023568]],  
      dtype=float32)>
```


Built model (simple)

```
[10]: model = tf.keras.Sequential()
model.add(hub_layer)
model.add(tf.keras.layers.Dense(16, activation='relu'))
model.add(tf.keras.layers.Dense(1))

model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
keras_layer (KerasLayer)	(None, 50)	48190600
dense (Dense)	(None, 16)	816
dense_1 (Dense)	(None, 1)	17

=====
Total params: 48,191,433
Trainable params: 48,191,433
Non-trainable params: 0
=====

Compile model

```
[11]: model.compile(optimizer='adam',
                    loss = tf.keras.losses.BinaryCrossentropy(from_logits=True),
                    metrics=['accuracy']
                    )
```

Run the model

```
[12]: history = model.fit(train_data.shuffle(10000).
    ↳ batch(512), epochs=10, validation_data=validation_data.batch(512), verbose=1)
```

Epoch 1/10

30/30 [=====] - 16s 511ms/step - loss: 0.6919 -
accuracy: 0.4962 - val_loss: 0.6285 - val_accuracy: 0.5625

Epoch 2/10

30/30 [=====] - 14s 484ms/step - loss: 0.5951 -
accuracy: 0.6127 - val_loss: 0.5218 - val_accuracy: 0.7233

Epoch 3/10

30/30 [=====] - 14s 483ms/step - loss: 0.4593 -
accuracy: 0.7734 - val_loss: 0.4134 - val_accuracy: 0.8097

Epoch 4/10

30/30 [=====] - 15s 485ms/step - loss: 0.3192 -
accuracy: 0.8775 - val_loss: 0.3525 - val_accuracy: 0.8446

Epoch 5/10

```
30/30 [=====] - 15s 484ms/step - loss: 0.2293 -  
accuracy: 0.9173 - val_loss: 0.3262 - val_accuracy: 0.8606  
Epoch 6/10  
30/30 [=====] - 15s 484ms/step - loss: 0.1676 -  
accuracy: 0.9477 - val_loss: 0.3139 - val_accuracy: 0.8601  
Epoch 7/10  
30/30 [=====] - 14s 484ms/step - loss: 0.1237 -  
accuracy: 0.9659 - val_loss: 0.3124 - val_accuracy: 0.8675  
Epoch 8/10  
30/30 [=====] - 15s 486ms/step - loss: 0.0878 -  
accuracy: 0.9775 - val_loss: 0.3177 - val_accuracy: 0.8656  
Epoch 9/10  
30/30 [=====] - 15s 486ms/step - loss: 0.0634 -  
accuracy: 0.9882 - val_loss: 0.3284 - val_accuracy: 0.8700  
Epoch 10/10  
30/30 [=====] - 15s 485ms/step - loss: 0.0453 -  
accuracy: 0.9926 - val_loss: 0.3411 - val_accuracy: 0.8677
```

[]: