Intro to Neural Nets

RNNs for Text

Today's Agenda

Background on NLP

- Use Cases
- Quick review on bag of words approaches, etc.

TextVectorization Layer

- This implements basic standardization and punctuation removal. It assumes 1-grams, then one-hot encodes.
- No stemming or stop word removal, by default.

Sequence vs. Bag-of-Words

Conceptually

Architectures for Sequences

Bidirectional LSTM



Quick Review of NLP Concepts

Pre-processing Text

- Lower-casing, stop word removal, stemming, removing punctuation, stripping rare tokens, etc.
- Tokenization (this may be chars, words, sentences, etc.
- Integer encoding / indexing the tokens.
- Finally, I may or may not leverage sequence information.
- Q: what is a bag of words approach? What are n-grams?

	Database	SQL	Index	Regression	Likelihood	linear
D1	24	21	9	0	0	3
D2	32	10	5	0	3	0
D3	12	16	5	0	0	0
D4	6	7	2	0	0	0
D5	43	31	20	0	3	0
D6	2	0	0	18	7	6
D7	0	0	1	32	12	0
D8	3	0	0	22	4	4
D9	1	0	0	34	27	25

Weighting Term-Documents: TF-IDF

Not all phrases are of equal importance...

- E.g., David less important than Beckham
- If a term occurs all the time, observing its presence is less informative

Inverse-document frequency (IDF) helps address this.

$$\mathrm{IDF} = \log(N/n_j)$$

- Term 'weighting' is then calculated as Term Frequency (TF) x IDF
- n_i = # of docs containing the term, N = total # of docs
- A term is deemed important if it has a high TF and/or a high IDF.
- As TF goes up, the word is more common generally. As IDF goes up, it means very few documents contain this term.

TextVectorization Layer

Pre-processing Text

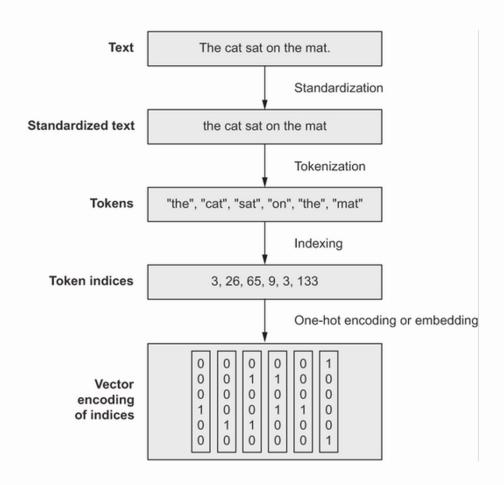
- Standardization, tokenization (words), one-hot-encoding / vectorization.
- The Keras TextVectorization() layer achieves these steps quickly.

Customization

 You can work with n-grams, and do other sorts of pre-processing, using arguments.

Options

- Include as part of TF Dataset pipeline (more efficient)
- Include as a layer in your Keras model.



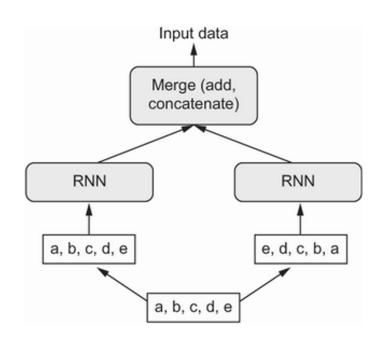
Bidirectional LSTM

We Saw This Last Time

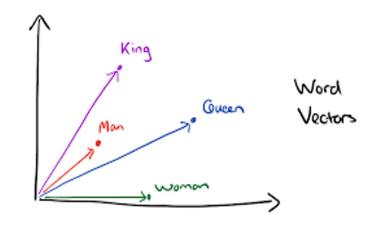
- Take each sequence as input data, as well as a flipped/reversed copy.
- Was state of the art for text processing until relatively recently (transformers now dominate).

Instead of Time Series We Pass...

- Sequences of one-hot-encodings of terms.
- Sequences of pre-trained vector embeddings of terms.



Embedding Layer



With Hot Encodings, Model Will Still Struggle to Figure Out Semantics

 Despite having sequence, the model is "told" that the tokens are orthogonal / independent of one another in their meanings. But that's not true!

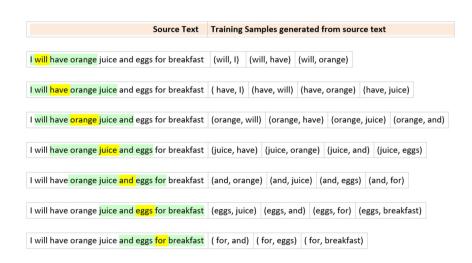
Textual Embedding Layer First Provides Dimensionality Reduction

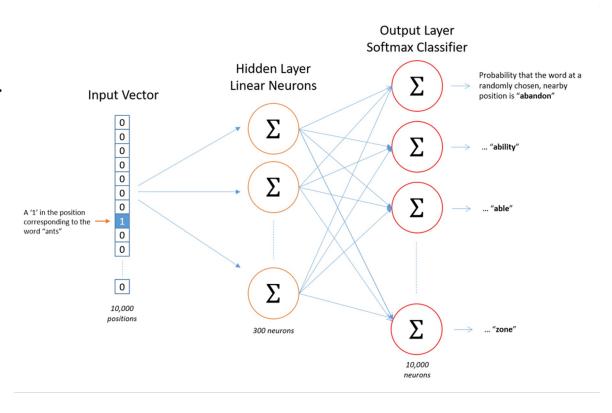
- Represent words into a lower dimensional space similar vector = similar meaning.
- The Embedding layer is a lookup table that maps tokens to vectors. For each token in the vocabulary, the network learns a vector representation. The vectors are initially random, and the network updates them in training to learn representations that help in prediction (just like with convolution filters!).
- In practice, it is learning semantic relationships...
- This is much better for an RNN than a hot encoding, because 120 values (for example) is << 20,000!

Pre-Trained Embeddings: Word2Vec

Word2Vec

- Two types: CBoW and Skipgram
- Construct training examples and labels.





Pre-Trained Embeddings: Limitation

Out of Sample Words

Both GloVe and Word2Vec are limited to words you've seen before in training.
 They cannot handle new words. Those words thus get omitted / dropped, or you need to do something different.

FastText

An extension to Word2Vec which learns character n-grams of words. So, instead
of embedding words, we embed portions of words (e.g., a 3-gram character
representation would break up the word 'coffee' into 'cof', 'off', 'ffe', ... and then
learn vector embeddings of each.



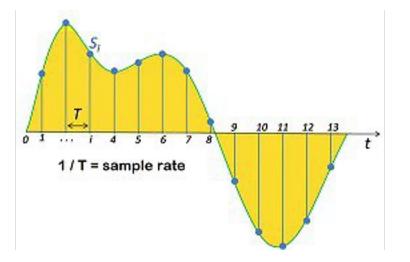
RNN for Audio

Same Sequence Concepts Work for Audio Data

 Audio files are just sequences of numeric values (amplitude), possibly two if it was recorded in stereo.

Once we recognize this, we realize we can predict things about audio sequences

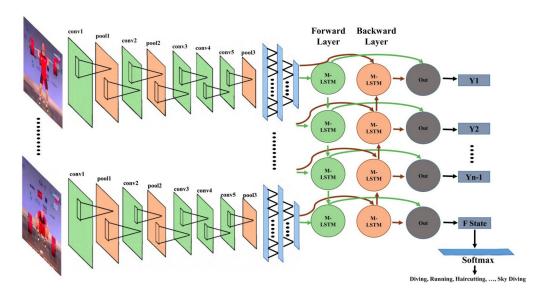
too!



CNN-RNN for Video

Hybrid Topology for Image Sequences

- We Use CNN's to detect features at a given input.
- We feed those feature maps into an RNN architecture, like LSTM.
- We can use this topology to predict things about videos.
- You might pre-process frames using a pre-trained CNN and pass feature maps as sequences to an RNN.



Questions?