Natural Language Processing Seq2seq for Text Generation



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- 1. Multiclass
- 2. Seq2seq
- 3. Task: Generate poems

Multiclass



Classifier Layer



- Final layer of a DNN: can be seen as a classifier
- Binary classification:

```
model.add(Dense(1, activation=tf.nn.sigmoid))
```

Multiclass: # class = num_class

```
model.add(Dense(num class, activation=tf.nn.softmax))
```

More Preprocessing



Label must be "one-hot"

Original labels: ["1", "2", ... "n"], e.g., n=5

One-hot label: [[1,0,0,0,0], [0,1,0,0,0], ..., [0,0,0,0,1]]

Keras has a helper:

```
tf.keras.utils.to categorical(labels, num classes=5)
```

- For Keras, labels must start from 0!
- Keep original labels for later

Redefine Loss



- Binary classification loss: 'binary_crossentropy'
- Multiclass classification loss: 'categorical_crossentropy'

[Source: Mikolov et al, NIPS'13]





- You may remember F-score
- However, F-score is NOT included in Keras
- Use sklearn, but remember sklearn requires original labels!

```
from sklearn.metrics import f1_score
y_true = [0, 1, 2, 3, 4, 2]
y_pred = [0, 2, 1, 4, 3, 1]
f1 score(y true, y pred, average=None)
```





Sequence to sequence

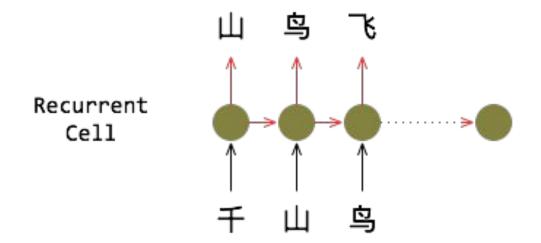


- Input and output are both sequences
- We can use this structure to do e.g., language modeling
 - Given previous words, predict the next word
 - Example task: generate Chinese poems
- A natural choice of neural network is RNN (more specifically, LSTM)
 - Output depends on previous input(s)
- However, as we mentioned, the sequence length cannot be too long

RNN for Language Modeling



Output is input shift by 1 time step







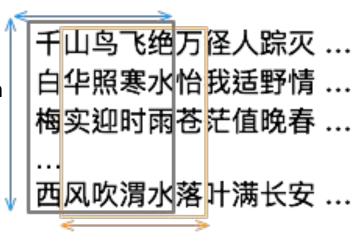
- You can download 全唐诗 from the web
- ~43,000 poems
- Need preprocessing like we mentioned before
- You can also add more custom data!

Preprocessing for LM training



- In practice:
 - Concatenate all poems into 1 vector
 - Reshape it to form a matrix
 - Take a part of the matrix as one batch

1 batch of input



1 batch of output





- Average loss of the entire sequence
 - o tf.contrib.seq2seq.sequence_loss
 - Weighted cross-entropy loss for a sequence
 - Can set mask for padding symbols (we won't be using this now)

Tensorflow Basics



Major steps



- Define hyperparameter
- Input generator
- Build model structure
- Training operations
- Training controller
- Testing operations

Major Object Types



- Operation (op): add, reshape, etc.
 - No gradients
- Variable: weights, embeddings
 - Have gradients (can be turned off)
- Placeholders: a way for you to assign values into the model
 - No gradients
- Session: like a thread that actually send and receive data to/from model

Main Concept



- Build "computation graph": define elements in the model and how they should be calculated/connected. Not actually running!
 - Caveat: It is very difficult to find bugs because the shape of the tensors are flexible. print does not work!
 - Includes calculating loss and gradients.
- Training controller creates data flow and records output from graph

Training Techniques



- During training, sometimes it's helpful to decay the learning rate over time
- Today, we will use the decay method after training for some epochs, or so-called "warmup" epoch.
- After warmup, we can decrease learning rate exponentially.

Code: Generate Chinese Poems

