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Introduction to Named Entity Recognition

What is Named Entity Recognition?

- Locates and extracts predefined entities from text
- Places, organizations, names, time and dates



Types of Entities







Thailand: Geographical

Google: Organization

Indian: Geopolitical

More Types of Entities



December: Time Indicator

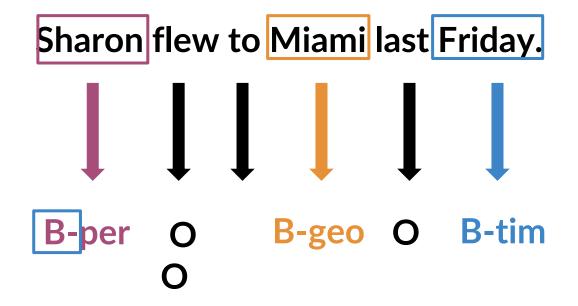


Egyptian statue: Artifact



Barack Obama: Person

Example of a labeled sentence



Applications of NER systems

- Search engine efficiency
- Recommendation engines
- Customer service
- Automatic trading









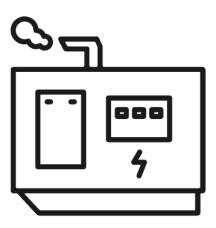


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Training NERs: Data Processing

Outline

- Convert words and entity classes into arrays
- Token padding
- Create a data generator



Processing data for NERs

- Assign each class a number
- Assign each word a number

per

Token padding

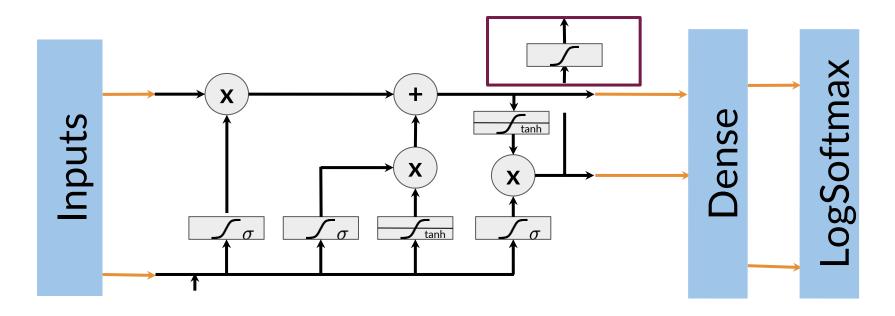
For LSTMs, all sequences need to be the same size.

- Set sequence length to a certain number
- Use the <PAD> token to fill empty spaces

Training the NER

- 1. Create a tensor for each input and its corresponding number
- 2. Put them in a batch 64, 128, 256, 512 ...
- 3. Feed it into an LSTM unit
- 4. Run the output through a dense layer
- 5. Predict using a log softmax over K classes

Training the NER



Layers in Trax

```
model = tl.Serial(
    tl.Embedding(),
    tl.LSTM(),
    tl.Dense()
    tl.LogSoftmax()
)
```

Summary

- Convert words and entities into same-length numerical arrays
- Train in batches for faster processing
- Run the output through a final layer and activation





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Computing Accuracy

Evaluating the model

- 1. Pass test set through the model
- 2. Get arg max across the prediction array
- 3. Mask padded tokens
- 4. Compare outputs against test labels

Evaluating the model in Python

```
def evaluate_model(test_sentences, test_labels, model):
    pred = model(test_sentences)
    outputs = np.argmax(pred, axis=2)
    mask = ...
    accuracy =
np.sum(outputs==test_labels)/float(np.sum(mask))
    return accuracy
```

Summary

- If padding tokens, remember to mask them when computing accuracy
- Coding assignment!