

# Join\_text\_and\_image\_representation

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## 1 Join text and image representation

```
[1]: import tensorflow
from tensorflow.keras.models import Sequential, load_model
import numpy as np
from tensorflow.keras.applications.resnet50 import ResNet50
from tensorflow.keras.applications.resnet50 import preprocess_input, \
    decode_predictions
from tensorflow.keras.preprocessing import image
from tensorflow.keras.applications.imagenet_utils import preprocess_input
[2]: resnet_model = ResNet50(weights='imagenet', include_top=False)

[3]: def extract_features(img_path):
    img = image.load_img(img_path, target_size=(224, 224)) x =
    image.img_to_array(img)
    x = np.expand_dims(x, axis=0) x =
    preprocess_input(x)
    features = resnet_model.predict(x)
    return np.expand_dims(features.flatten(), axis=0)
```

## 2 Image representation

```
[60]: features = np.load('resnet50-features.10k.npy')
print(features.shape)
```

(10000, 2048)

```
[ ]:
```

### 3 Text Representation

```
[4]: def load(captions_filename, features_filename): features =  
    np.load(features_filename)  
    images = [] texts = []  
    with open(captions_filename) as fp:  
        for line in fp:  
            tokens = line.strip().split()  
            images.append(tokens[0]) texts.append(''  
                '.join(tokens[1:]))  
    return features, images, texts
```

```
[5]: feature , images , texts = load( "annotations.10k.txt", "resnet50-features.10k."  
    ↪ "npz")
```

```
[6]: from keras.preprocessing.text import Tokenizer  
    from keras.preprocessing.sequence import pad_sequences  
    tokenizer = Tokenizer()  
    tokenizer.fit_on_texts(texts)  
    sequences = tokenizer.texts_to_sequences(texts)  
    captions = pad_sequences(sequences, maxlen=16)
```

```
[7]: vocab = tokenizer.word_index  
    vocab['<eos>'] = 0
```

```
[8]: import json  
    with open('vocab.json', 'w') as fp: # save the vocab  
        fp.write(json.dumps(vocab))
```

#### 3.1 load the pretrained embedded model

```
[9]: import embedding  
    embedding_weights = embedding.load(vocab, 100, 'glove.twitter.27B.100d.filtered.'  
    ↪ ".txt")
```

loading embeddings from "glove.twitter.27B.100d.filtered.txt"

### 4 putting together the model

```
[10]: from keras.layers import Input, Dense, Embedding, GRU  
    image_input = Input(shape=(2048,))  
    caption_input = Input(shape=(16,)) noise_input =  
    Input(shape=(16,))
```

```
[11]: caption_embedding = Embedding(len(vocab), 100,
    ↪input_length=16, weights=[embedding_weights])
caption_rnn = GRU(256)
image_dense = Dense(256, activation='tanh')
```

#### 4.1 creation of pipe line

```
[12]: image_pipeline = image_dense(image_input)
caption_pipeline = caption_rnn(caption_embedding(caption_input))
noise_pipeline = caption_rnn(caption_embedding(noise_input))
```

#### 4.2 compute the dot product between image and caption

```
[13]: from keras.layers import merge
positive_pair = merge.dot([image_pipeline, caption_pipeline] , axes=1,
    ↪normalize=False )
negative_pair = merge.dot([image_pipeline, noise_pipeline], axes=1,
    ↪normalize=False)
output = merge.concatenate([positive_pair, negative_pair] )
```

```
[14]: positive_pair, negative_pair, output
```

```
[14]: (<tf.Tensor 'dot/Squeeze:0' shape=(None, 1) dtype=float32>,
    <tf.Tensor 'dot_1/Squeeze:0' shape=(None, 1) dtype=float32>,
    <tf.Tensor 'concatenate/concat:0' shape=(None, 2) dtype=float32>)
```

#### 4.3 create multiple models

```
[15]: from tensorflow.keras import Model
training_model = Model(inputs=[image_input, caption_input, noise_input],
    ↪outputs=output)
image_model = Model(inputs=image_input, outputs=image_pipeline) caption_model =
Model(inputs=caption_input, outputs=caption_pipeline)
```

### 5 Custom loss

```
[16]: from keras import backend as K
def custom_loss(y_true, y_pred): positive
    = y_pred[:,0] negative = y_pred[:,1]
    return K.sum(K.maximum(0., 1. - positive + negative))
```

```
[17]: def accuracy(y_true, y_pred): positive =
    y_pred[:,0] negative = y_pred[:,1]
```

```
return K.mean(positive > negative)
```

```
[18]: training_model.compile(loss=custom_loss,optimizer='adam',metrics=[accuracy])
```

## 6 Training

### 6.1 spiting the model

```
[61]: noise = np.copy(captions)
fake_labels = np.zeros((len(features), 1))
X_train = ( [features[:9000], captions[:9000], noise[:9000]]) Y_train = (
fake_labels[:9000])
X_valid = [features[-1000:], captions[-1000:],noise[-1000:]] Y_valid =
fake_labels[-1000:]
```

```
[62]: ### actual trainig
```

```
[63]: for epoch in range(10):
    np.random.shuffle(noise) # don't forget to shuffle mismatched captions
    training_model.fit(X_train, Y_train,validation_data=[X_valid, Y_valid],
    ↪epochs =1,batch_size=64)
```

141/141[=====]	- 20s 141ms/step - loss: 4.3636 -
accuracy: 0.9726 - val_loss: 0.0000e+00	- val_accuracy: 0.0000e+00
141/141[=====]	- 20s 142ms/step - loss: 4.5233 -
accuracy: 0.9720 - val_loss: 0.0000e+00	- val_accuracy: 0.0000e+00
141/141[=====]	- 20s 141ms/step - loss: 4.1314 -
accuracy: 0.9740 - val_loss: 0.0000e+00	- val_accuracy: 0.0000e+00
141/141[=====]	- 20s 141ms/step - loss: 4.4201 -
accuracy: 0.9725 - val_loss: 0.0000e+00	- val_accuracy: 0.0000e+00
141/141[=====]	- 22s 153ms/step - loss: 4.4955 -
accuracy: 0.9735 - val_loss: 0.0000e+00	- val_accuracy: 0.0000e+00
141/141[=====]	- 20s 142ms/step - loss: 4.3632 -
accuracy: 0.9743 - val_loss: 0.0000e+00	- val_accuracy: 0.0000e+00
141/141[=====]	- 20s 143ms/step - loss: 3.9072 -
accuracy: 0.9767 - val_loss: 0.0000e+00	- val_accuracy: 0.0000e+00
141/141[=====]	- 20s 142ms/step - loss: 3.9071 -
accuracy: 0.9771 - val_loss: 0.0000e+00	- val_accuracy: 0.0000e+00
141/141[=====]	- 20s 142ms/step - loss: 3.8084 -
accuracy: 0.9769 - val_loss: 0.0000e+00	- val_accuracy: 0.0000e+00
141/141[=====]	- 20s 141ms/step - loss: 3.9775 -
accuracy: 0.9743 - val_loss: 0.0000e+00	- val_accuracy: 0.0000e+00

## 6.2 saving model and representation

```
[64]: image_model.save('model.image') caption_model.save('model.caption')

np.save('caption-representations',caption_model.predict(captions)) np.save('image-
representations', image_model.predict(feature))
```

INFO:tensorflow:Assets written to: model.image/assets

INFO:tensorflow:Assets written to: model.caption/assets

## 7 Captioning novel images

```
[65]: from keras.models import load_model
image_model = load_model('model.image',compile=False)
caption_model =load_model('model.caption',compile=False)
```

### 7.1 load representation

```
[66]: import numpy as np
caption_representations =np.load('caption-representations.npy')
image_representations = np.load('image-representations.npy')
```

```
[67]: from keras.preprocessing.sequence import pad_sequences
import json
vocab = json.loads(open('vocab.json').read())
def preprocess_texts(texts): output = []
    for text in texts:
        output.append([vocab[word] if word in vocab else 0 for word in text.
↪split()])
    return pad_sequences(output, maxlen=16)
```

```
[68]: def generate_caption(image_filename, n=10):
    # generate image representation for new image
    image_representation = image_model.predict(extract_features(image_filename))
    # compute score of all captions in the dataset
    scores = np.dot(caption_representations,image_representation.T).flatten()
    # compute indices of n best captions
    indices = np.argpartition(scores, -n)[-n:]

    indices = indices[np.argsort(scores[indices])]
    # display them
    for i in [int(x) for x in reversed(indices)]: print(scores[i],
        texts[i])
```

```
[75]: #generate_caption('images.jpg')
```

## 8 searching for images

```
[70]: def search_image(caption, n=10):  
      caption_representation = caption_model.predict(preprocess_texts([caption])) scores =  
      np.dot(image_representations, caption_representation.T).flatten() indices =  
      np.argpartition(scores, -n)[-n:]  
      indices = indices[np.argsort(scores[indices])]   
      for i in [int(x) for x in reversed(indices)]: print(scores[i],  
              images[i])
```

```
[71]: search_image('a man in the snow on some skis')
```

```
15.56558 COCO_val2014_000000022626.jpg  
13.151729 COCO_val2014_000000007938.jpg  
11.980955 COCO_val2014_0000000276893.jpg  
11.921771 COCO_val2014_0000000380906.jpg  
11.8630085 COCO_val2014_0000000392575.jpg  
11.753822 COCO_val2014_000000012946.jpg  
11.510327 COCO_val2014_0000000134343.jpg  
11.507981 COCO_val2014_0000000119773.jpg  
11.371106 COCO_val2014_0000000204360.jpg  
11.25193 COCO_val2014_0000000112769.jpg
```

```
[71]:
```