```
In [ ]:
In [11]: | # As usual, a bit of setup
         from future import print function
         import time, os, json
         import numpy as np
         import matplotlib.pyplot as plt
         from deeplearning.gradient check import eval numerical gradient, eval
          numerical gradient array
         from deeplearning.rnn layers import *
         from deeplearning.captioning solver import CaptioningSolver
         from deeplearning.classifiers.rnn import CaptioningRNN
         from deeplearning.coco utils import load coco data, sample coco minib
         atch, decode captions
         from deeplearning.image utils import image from url
         %matplotlib inline
         plt.rcParams['figure.figsize'] = (10.0, 8.0) # set default size of pl
         plt.rcParams['image.interpolation'] = 'nearest'
         plt.rcParams['image.cmap'] = 'gray'
         # for auto-reloading external modules
         # see http://stackoverflow.com/questions/1907993/autoreload-of-module
         s-in-ipython
         %load ext autoreload
         %autoreload 2
         def rel error(x, y):
             """ returns relative error """
             return np.max(np.abs(x - y) / (np.maximum(1e-8, np.abs(x) + np.ab)
         s(y))))
```

Install h5py

The COCO dataset we will be using is stored in HDF5 format. To load HDF5 files, we will need to install the h5py Python package. From the command line, run: pip install h5py

If you receive a permissions error, you may need to run the command as root: sudo pip install h5py

You can also run commands directly from the Jupyter notebook by prefixing the command with the "!" character:

In [10]: |!pip install h5py

Requirement already satisfied: h5py in ./.env/lib/python3.6/site-pack ages (2.7.0)
Requirement already satisfied: six in ./.env/lib/python3.6/site-packa ges (from h5py) (1.10.0)
Requirement already satisfied: numpy>=1.7 in ./.env/lib/python3.6/sit e-packages (from h5py) (1.12.1)

```
In [12]: # Load COCO data from disk; this returns a dictionary
    # We'll work with dimensionality-reduced features for this notebook,
    but feel
    # free to experiment with the original features by changing the flag
    below.
    data = load_coco_data(pca_features=True)

# Print out all the keys and values from the data dictionary
for k, v in data.items():
    if type(v) == np.ndarray:
        print(k, type(v), v.shape, v.dtype)
    else:
        print(k, type(v), len(v))
```

train_captions <class 'numpy.ndarray'> (400135, 17) int32 train_image_idxs <class 'numpy.ndarray'> (400135,) int32 val_captions <class 'numpy.ndarray'> (195954, 17) int32 val_image_idxs <class 'numpy.ndarray'> (195954,) int32 train_features <class 'numpy.ndarray'> (82783, 512) float32 val_features <class 'numpy.ndarray'> (40504, 512) float32 idx_to_word <class 'list'> 1004 word_to_idx <class 'dict'> 1004 train_urls <class 'numpy.ndarray'> (82783,) <U63 val urls <class 'numpy.ndarray'> (40504,) <U63

```
In [13]: # Sample a minibatch and show the images and captions
batch_size = 3

captions, features, urls = sample_coco_minibatch(data, batch_size=bat ch_size)
for i, (caption, url) in enumerate(zip(captions, urls)):
    plt.imshow(image_from_url(url))
    plt.axis('off')
    caption_str = decode_captions(caption, data['idx_to_word'])
    plt.title(caption_str)
    plt.show()
```

<START> a woman wearing a short <UNK> kneeling on a tennis court <END>





<START> a tour bus driving through an empty intersection <END>



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<START> a person skiing on skis on a snow covered hill <END>



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```
np.random.seed(231)
In [24]:
         small_data = load_coco_data(max_train=50)
         small rnn model = CaptioningRNN(
                    cell_type='rnn',
                    word to idx=data['word to idx'],
                    input_dim=data['train_features'].shape[1],
                    hidden dim=512,
                    wordvec_dim=256,
         small rnn solver = CaptioningSolver(small rnn model, small data,
                     update rule='adam',
                     num epochs=50,
                     batch_size=25,
                     optim config={
                       'learning_rate': 5e-3,
                     },
                     lr decay=0.95,
                     verbose=True, print every=10,
         small rnn solver.train()
         # Plot the training losses
         plt.plot(small rnn solver.loss history)
         plt.xlabel('Iteration')
         plt.ylabel('Loss')
         plt.title('Training loss history')
         plt.show()
```

```
(Iteration 1 / 100) loss: 76.913487

(Iteration 11 / 100) loss: 21.062919

(Iteration 21 / 100) loss: 4.016281

(Iteration 31 / 100) loss: 0.566975

(Iteration 41 / 100) loss: 0.239450

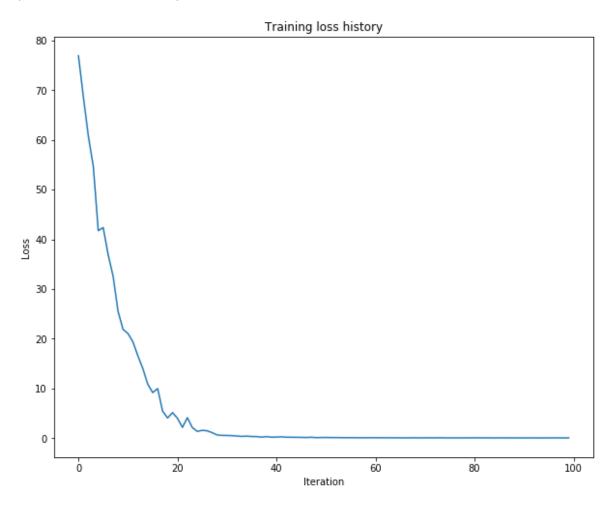
(Iteration 51 / 100) loss: 0.161949

(Iteration 61 / 100) loss: 0.111534

(Iteration 71 / 100) loss: 0.097569

(Iteration 81 / 100) loss: 0.099074

(Iteration 91 / 100) loss: 0.073962
```



Test-time sampling

train
a closeup of a woman from the <UNK> to <UNK> <END>
GT:<START> a closeup of a woman from the <UNK> to <UNK> <END>



train
a man standing on the side of a road with bags of luggage <END>
GT:<START> a man standing on the side of a road with bags of luggage <END>



val sitting atop the pile of skiing gear in room <UNK> with wood <END> GT:<START> a man looks <UNK> and holds up his kite <END>



val a dog is on the <END> GT:<START> a white bed in the middle of a home room <END>



In	[1:	
In	[1:	