

softmax

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1 Softmax Classifier

This exercise guides you through the process of classifying images using a Softmax classifier. As part of this you will:

- Implement a fully vectorized loss function for the Softmax classifier
- Calculate the analytical gradient using vectorized code
- Tune hyperparameters on a validation set
- Optimize the loss function with Stochastic Gradient Descent (SGD)
- Visualize the learned weights

```
[80]: # start-up code!
import random

import matplotlib.pyplot as plt
import numpy as np

%matplotlib inline
plt.rcParams['figure.figsize'] = (10.0, 8.0) # set default size of plots
plt.rcParams['image.interpolation'] = 'nearest'
plt.rcParams['image.cmap'] = 'gray'

# for auto-reloading extenrnal modules
# see http://stackoverflow.com/questions/1907993/
  ↪ autoreload-of-modules-in-ipython
%load_ext autoreload
%autoreload 2
```

The autoreload extension is already loaded. To reload it, use:

```
%reload_ext autoreload
```

```
[81]: from load_cifar10_tvt import load_cifar10_train_val

X_train, y_train, X_val, y_val, X_test, y_test = load_cifar10_train_val()
print("Train data shape: ", X_train.shape)
print("Train labels shape: ", y_train.shape)
print("Val data shape: ", X_val.shape)
print("Val labels shape: ", y_val.shape)
```

```
print("Test data shape: ", X_test.shape)
print("Test labels shape: ", y_test.shape)
```

Train, validation and testing sets have been created as

X_i and y_i where $i=\text{train, val, test}$

Train data shape: (3073, 49000)

Train labels shape: (49000,)

Val data shape: (3073, 1000)

Val labels shape: (1000,)

Test data shape: (3073, 1000)

Test labels shape: (1000,)

Code for this section is to be written in `cs231n/classifiers/softmax.py`

```
[156]: # Now, implement the vectorized version in softmax_loss_vectorized.

import time

from cs231n.classifiers.softmax import softmax_loss_vectorized

# gradient check.
from cs231n.gradient_check import grad_check_sparse

W = np.random.randn(10, 3073) * 0.0001

tic = time.time()
loss, grad = softmax_loss_vectorized(W, X_train, y_train, 0.00001)
toc = time.time()
print("vectorized loss: %e computed in %fs" % (loss, toc - tic))

# As a rough sanity check, our loss should be something close to -log(0.1).
print("loss: %f" % loss)
print("sanity check: %f" % (-np.log(0.1)))

f = lambda w: softmax_loss_vectorized(w, X_train, y_train, 0.0)[0]
grad_numerical = grad_check_sparse(f, W, grad, 10)
```

vectorized loss: 2.317796e+00 computed in 0.984168s

loss: 2.317796

sanity check: 2.302585

numerical: 2.060581 analytic: 2.060581, relative error: 8.199045e-09

numerical: -1.910819 analytic: -1.910819, relative error: 4.206250e-09

numerical: 0.747108 analytic: 0.747108, relative error: 2.444263e-08

numerical: 0.396134 analytic: 0.396134, relative error: 5.135847e-09

numerical: 1.727855 analytic: 1.727855, relative error: 5.932070e-08

numerical: 0.781361 analytic: 0.781361, relative error: 8.113437e-09

numerical: -2.858721 analytic: -2.858721, relative error: 8.045210e-09

numerical: 1.416608 analytic: 1.416608, relative error: 2.640299e-08

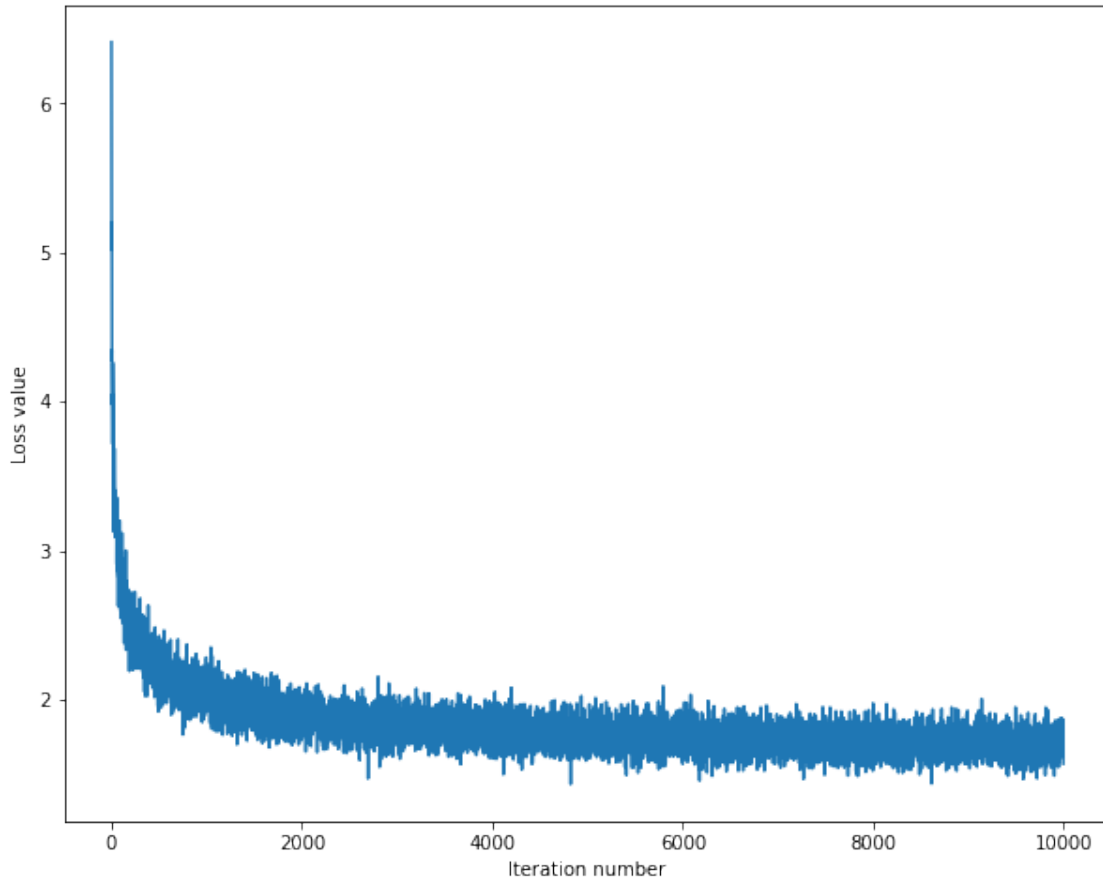
numerical: -3.852705 analytic: -3.852705, relative error: 1.780070e-09
numerical: 0.527616 analytic: 0.527616, relative error: 1.419871e-08

Code for this section is to be written in `cs231n/classifiers/linear_classifier.py`

```
[157]: # Now that efficient implementations to calculate loss function and gradient of  
↪ the softmax are ready,  
# use it to train the classifier on the cifar-10 data  
# Complete the `train` function in cs231n/classifiers/linear_classifier.py  
  
from cs231n.classifiers.linear_classifier import Softmax  
  
classifier = Softmax()  
loss_hist = classifier.train(  
    X_train,  
    y_train,  
    learning_rate=1e-6,  
    reg=1e-1,  
    num_iters=10000,  
    batch_size=200,  
    verbose=False,  
)  
  
# Training Accuracy  
y_train_pred = classifier.predict(X_train)  
train_accuracy = np.mean(y_train == y_train_pred)  
print("softmax on raw pixels training set accuracy: %f" % (train_accuracy,))  
  
# Plot loss vs. iterations  
plt.plot(loss_hist)  
plt.xlabel("Iteration number")  
plt.ylabel("Loss value")
```

softmax on raw pixels training set accuracy: 0.424755

```
[157]: Text(0, 0.5, 'Loss value')
```



```
[158]: # Complete the `predict` function in cs231n/classifiers/linear_classifier.py
# Evaluate on test set
y_test_pred = classifier.predict(X_test)
test_accuracy = np.mean(y_test == y_test_pred)
print("softmax on raw pixels final test set accuracy: %f" % (test_accuracy,))
```

softmax on raw pixels final test set accuracy: 0.380000

```
[159]: # Visualize the learned weights for each class
w = classifier.W[:, :-1] # strip out the bias
w = w.reshape(10, 32, 32, 3)

w_min, w_max = np.min(w), np.max(w)

classes = [
    "plane",
    "car",
    "bird",
    "cat",
```

```

    "deer",
    "dog",
    "frog",
    "horse",
    "ship",
    "truck",
]
for i in range(10):
    plt.subplot(2, 5, i + 1)

    # Rescale the weights to be between 0 and 255
    wimg = 255.0 * (w[i].squeeze() - w_min) / (w_max - w_min)
    plt.imshow(wimg.astype("uint8"))
    plt.axis("off")
    plt.title(classes[i])

```

