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import numpy as np
import h5py
import pickle
import scipy
from scipy import ndimage
def load_data():
Load Data Function and training data are taken from coursera's machine learning course.
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  train_dataset = h5py.File('datasets/train_catvnoncat.h5', "r")
  train_x = np.array(train_dataset["train_set_x"][:])
  train_y = np.array(train_dataset["train_set_y"][:])
  test_dataset = h5py.File('datasets/test_catvnoncat.h5', "r")
  test_x = np.array(test_dataset["test_set_x"][:])
  test_y = np.array(test_dataset["test_set_y"][:])
  train_y = train_y.reshape((1, train_y.shape[0]))
  test_y = test_y.reshape((1, test_y.shape[0]))
  return train_x, train_y, test_x, test_y
def save_obj(obj, name):
  with open(name + '.pkl', 'wb') as f:
    pickle.dump(obj, f, pickle.HIGHEST_PROTOCOL)
    print("Saved file")
```

```
def sigmoid(Z):
  return 1/(1 + np.exp(-Z))
def initialize(dimensions):
  w = np.zeros(shape=(dimensions, 1))
  b = 0
  assert(w.shape == (dimensions, 1))
  assert(isinstance(b, float) or isinstance(b, int))
  return w, b
def cost(A, Y, x_shape):
  return (-1/x_{p,-1})^*np.sum(Y^*np.log(A) + (1 - Y)^*(np.log(1 - A)))
def forward_propagation(w, b, X, Y):
  x_shape= X.shape[1]
  A = sigmoid(np.dot(w.T, X) + b)
  return np.squeeze(cost(A, Y, x_shape)), A
def backward_propagation(w, b, X, A, Y):
  x_shape = X.shape[1]
  dw = (1 / x_shape) * np.dot(X, (A - Y).T)
  db = (1 / x_shape) * np.sum(A - Y)
  return dw, db
```

```
def gradient_descent(w, b, X, Y, iterations, alpha):
  costs = []
  for i in range(iterations):
    cost, A = forward_propagation(w, b, X, Y)
    dw, db = backward_propagation(w, b, X, A, Y)
    w = w- alpha * dw
    b = b- alpha * db
    if i % 100 == 0:
      costs.append(cost)
       print ("The cost after {} steps is: {}".format(i, cost))
  return w, b, dw, db, costs
def predict(w, b, X):
  x_shape = X.shape[1]
  prediction = np.zeros((1, x_shape))
  w = w.reshape(X.shape[0], 1)
  A = np.dot(w.T, X)+b
  A = sigmoid(A)
  for i in range(A.shape[1]):
```

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A[0, i] = 1 if A[0,i] > 0.5 else 0
```

return A

```
def predict_accuracy(X_train, Y_train, X_test, Y_test, parameters):
  w = parameters["w"]
  b = parameters["b"]
  test_prediction = predict(w, b, X_test)
  train_prediction = predict(w, b, X_train)
  test_accuracy = 100 - (np.mean(np.abs(test_prediction - Y_test)) * 100)
  train_accuracy = 100 - (np.mean(np.abs(train_prediction - Y_train)) * 100)
  return test_accuracy, train_accuracy
def model(x_train, y_train, iterations=2000, alpha=0.005):
  w, b = initialize(x_train.shape[0])
  w, b, dw, db, costs = gradient_descent(w, b, x_train, y_train, iterations, alpha)
  print("Model finished")
  parameters = {"w": w, "b" : b, "costs" : costs, "alpha" : alpha, "iterations": iterations}
  return parameters
```

```
x_train, y_train, x_test, y_test = load_data()
num_px = x_train.shape[1]
x_train = x_train.reshape(x_train.shape[0], -1).T
x_{test} = x_{test.reshape}(x_{test.shape}[0], -1).T
print("Reshaped train data of size: " + str(x_train.shape[0]))
print("Reshaped test data of size: " + str(x_test.shape[0]) + '\n')
x_{train} = x_{train} / 255.
x_{test} = x_{test} / 255.
parameters = model(x_train, y_train, iterations = 2000, alpha = 0.005)
test_accuracy, train_accuracy = predict_accuracy(x_train, y_train, x_test, y_test, parameters)
print("The train set had an accuracy of: {}".format(train_accuracy))
print("The test set had an accuracy of: {}".format(test_accuracy))
save_obj(parameters, "datasets/parameters")
```

```
# Test your own image
image = "image.jpg"

image = np.array(ndimage.imread(image, flatten = False))
image = scipy.misc.imresize(image, size=(num_px, num_px)).reshape((1, num_px ** 2 * 3)).T

numprediction = predict(parameters["w"], parameters["b"], image)
prediction = "Not a cat" if int(np.squeeze(numprediction)) == 0 else "A cat"

print("The algorithm predicts: " + prediction, np.squeeze(numprediction))
```