

Python vs. Go

A machine learning experiment



Chiara Mezzavilla Berlin Python Pizza August 2019

Hello!

I'm Chiara:-)

- Coding in Python since 2012
- Coaching at OpenTechSchool since 2013
- Paid developer since 2015
- Started learning Go to have a better grasp of CS concepts
- First time conference speaker!

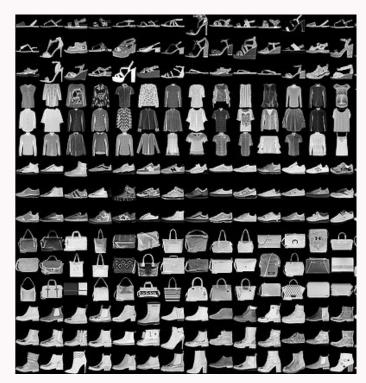
How do Python and Go compare, in classification?

Objectives:

- Get to know the Go machine learning stack (conference)
- Compare ease of use and performance with Python
- Try the Fashion MNIST dataset (contributed to by a former colleague)
- Give a talk at work :-)

Fashion MNIST replaces the digits MNIST dataset

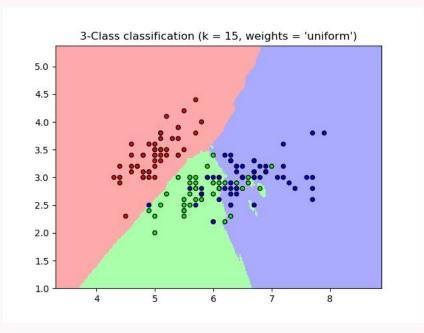
- 60000 records in the training dataset,
 10000 in the test dataset (but I used a smaller one to speed up)
- Images are represented by pixel color (greyscale)
- 10 labels, corresponding to categories of clothing, shoes, and accessories (t-shirts, pullovers, ankle boots, bags, etc.)
- Replaces the digits datasets, considered "too easy"



Source: https://github.com/zalandoresearch/fashion-mnist

KNN is a simple classification algorithm

- KNN looks at the "neighbors" to predict the label
- Easier to understand in a 2-d representation like that --->
- Doesn't make assumption on the distribution of data
- Doesn't do any actual learning, simply saves the training data data and compares the test data with it



Source: scikit-learn.org

First finding (surprise!): Go was easier to write

- It was tricky to load the data because of different assumptions
- Golearn provides a function to load data, which assumes that the rightmost column is the label
- Scikit-learn accepts numpy arrays you need to parse the csv files properly (unless one loads a built-in dataset)

```
main.go
                                                                                                   py_fashion_ml.py
package main
                                                                                                import numpy as np
                                                                                                from sklearn.neighbors import KNeighborsClassifier
import (
                                                                                                from sklearn.metrics import classification_report
  "fmt"
  "github.com/pkg/profile"
                                                                                                print("Loading data...")
                                                                                                train_data = np.loadtxt(fname='data/fashion-mnist_train.csv',
  "github.com/sjwhitworth/golearn/base"
  "github.com/sjwhitworth/golearn/evaluation"
                                                                                                                        delimiter=',',
  "github.com/sjwhitworth/golearn/knn"
                                                                                                                        skiprows=1)
                                                                                                test_data = np.loadtxt(fname='data/fashion-mnist_testsmall.csv',
                                                                                                                       delimiter=',',
func main() {
                                                                                                                       skiprows=1)
                                                                                                print("Very data, much mnist, wow")
  fmt.Println("Loading data...")
  trainData, err := base.ParseCSVToInstances("data/fashion-mnist_train.csv", true)
  if err != nil {
                                                                                                neigh = KNeighborsClassifier(n_neighbors=5, algorithm='kd_tree')
    panic(err)
                                                                                                print("Many training...")
  testData, err := base.ParseCSVToInstances("data/fashion-mnist_testsmall.csv", true)
                                                                                                neigh.fit(train_data[:, :-1], train_data[:, -1])
  if err != nil {
    panic(err)
                                                                                                print("Very predict...")
                                                                                                prediction = neigh.predict(test_data[:, :-1])
  fmt.Println("Very data, much mnist, wow")
                                                                                                print("Classification summary")
  classifier := knn.NewKnnClassifier("euclidean", "kdtree", 5)
                                                                                                print(classification_report(y_pred=prediction, y_true=test_data[:, -1]))
  fmt.Println("Many training...")
  err = classifier.Fit(trainData)
  if err != nil {
    panic(err)
  fmt.Println("Very predict...")
  predictions, err := classifier.Predict(testData)
  if err != nil {
  confusionMat, err := evaluation.GetConfusionMatrix(testData, predictions)
  if err != nil {
    panic(fmt.Sprintf("Unable to get confusion matrix: %s", err.Error()))
  fmt.Println(evaluation.GetSummary(confusionMat))
```

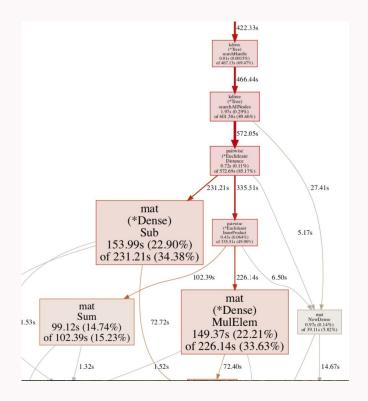
Second finding: accuracy was similar

```
time ./main
Loading data...
Very data, much mnist, wow
Many training...
Very predict...
                        False
                                            Precision Recall F1 Score
Reference
            True
                                  True
Classn
            Positives
                        Positives Negatives
                                  443
                                            0.9273
                                                       0.9808 0.9533
            45
                                  449
                                            0.9375
                                                       0.9574
                                                              0.9474
                        15
            41
                                  431
                                            0.7321
                                                       0.7736 0.7523
                                  452
                                            0.9091
                                                       0.9302 0.9195
            40
            45
                                  437
                                            0.8654
                                                       0.8182 0.8411
                                            0.6296
                                                              0.6602
            34
                        20
                                  430
                                                       0.6939
            45
                                  446
                                            0.9783
                                                       0.8654
                                                              0.9184
            39
                                            1.0000
                                                       0.9750
                                  459
                                                              0.9873
            50
                                  427
                                            0.8475
                                                       0.7937 0.8197
            41
                                            0.8913
                                                       0.9111 0.9011
                                  449
Overall accuracy: 0.8637
./main 554.05s user 2.83s system 103% cpu 9:00.66 total
```

```
time python py_fashion_ml.py
Loading data...
Very data, much mnist, wow
Many training...
Very predict...
Classification summary
              precision
                            recall f1-score
                                                support
         0.0
                    0.75
                              0.84
                                         0.79
                                                     98
         1.0
                    1.00
                              0.96
                                         0.98
                                                     89
         2.0
                    0.76
                              0.82
                                         0.79
                                                    111
         3.0
                    0.90
                              0.90
                                         0.90
                                                    104
                    0.80
                              0.79
                                         0.80
                                                     96
                    0.99
                              0.86
                                         0.92
         5.0
                                                     96
         6.0
                    0.68
                              0.60
                                         0.64
                                                    100
         7.0
                    0.91
                              0.98
                                         0.94
                                                    102
         8.0
                    0.95
                              0.94
                                         0.94
                                                     98
         9.0
                    0.93
                              0.95
                                         0.94
                                                    105
                                         0.86
                                                    999
    accuracy
                    0.87
                              0.86
                                         0.86
                                                    999
   macro avo
                                         0.86
                                                    999
weighted avg
                    0.87
                              0.86
python py_fashion_ml.py 105.02s user 1.67s system
99% cpu 1:46.78 total
```

Third finding (surprise?): C makes Python faster





Thank you!

Find me on Twitter @ChiaraM_87

See also:

- Scikit-learn https://github.com/scikit-learn/scikit-learn
- Golearn https://github.com/sjwhitworth/golearn
- Fashion MNIST https://github.com/zalandoresearch/fashion-mnist
- My code https://github.com/samurang87/go-fashion-ml