6.047 Problem Set 4 Writeup

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1 Simulated GWAS

- (a) β_i and SNP odds ratio
- (b)

See simulated_gwas.py.

- (c)
- (d)
- (e) Hyperparameters

$$\textbf{1.0.1} \quad n=10000, k=100, m=1000, s=0.25$$

 test

1.0.2
$$n = 1000, k = 100, m = 1000, s = 0.25$$

1.0.3
$$n = 100000, k = 100, m = 1000, s = 0.25$$

1.0.4
$$n = 10000, k = 100, m = 10000, s = 0.25$$

1.0.5
$$n = 10000, k = 100, m = 1000, s = 0.5$$

1.0.6
$$n = 10000, k = 100, m = 1000, s = 0.1$$

2 Finding eQTLs

- (a) Principle Components Analysis
- (b) Finding eQTLs via Linear Regression

For every SNP x_i , we find the mean and variance (μ_i, σ_i^2) of the correlation coefficients r_{ij} that x_i has with expression of gene y_j . In this way, we are determining the "typical" contribution of SNP x_i to any gene. We then select the SNP and gene pairs (x_i, e_j) for which r_{ij} is satistically significant under the null hypothesis $H_0: \rho_{ij} = \mu_i$ (i.e. contribution of x_i to y_j is typical).

To implement the Bonferroni correction, we test each hypothesis that SNP x_i contributes to the expression of gene y_j with significance of $\alpha/5000$, where $\alpha = 0.05$ (since we are testing 5000 different genes corresponding to 5000 different hypotheses).

(c) Additional Datasets

3 Convolutional Neural Networks

(a) Implementation

```
#!/usr/bin/env python
from keras.models import *
from keras.layers import *
import keras
import numpy as np
from datetime import datetime
import alternative_models as models
import argparse
BATCH_SIZE = 10
NUM_EPOCHS = 20
KERNEL_SIZE = (4, 4)
POOL_SIZE = (4, 6)
HIDDEN_UNITS = 32
CONV_FILTERS = 32
MODEL_NAME = None
def get_x_y_data():
   negative_data = []
   with open('negativedata.txt') as f:
        for line in f:
            final_mat = np.zeros((4,len(line)-1,1))
            for i in range(len(line)):
                char = line[i]
                if char == 'a':
                    final_mat[:,i,:] = np.array([[1],[0],[0],[0]])
                if char == 'c':
                    final_mat[:,i,:] = np.array([[0],[1],[0],[0]])
                if char == 'g':
                    final_mat[:,i,:] = np.array([[0],[0],[1],[0]])
```

```
if char == 't':
                    final_mat[:,i,:] = np.array([[0],[0],[0],[1]])
            negative_data.append(final_mat)
    positive_data = []
    with open('positivedata.txt') as f:
        for line in f:
            final_mat = np.zeros((4,len(line)-1,1))
            for i in range(len(line)):
                char = line[i]
                if char == 'a':
                    final_mat[:,i,:] = np.array([[1],[0],[0],[0]])
                if char == 'c':
                    final_mat[:,i,:] = np.array([[0],[1],[0],[0]])
                if char == 'g':
                    final_mat[:,i,:] = np.array([[0],[0],[1],[0]])
                if char == 't':
                    final_mat[:,i,:] = np.array([[0],[0],[0],[1]])
            positive_data.append(final_mat)
   X = np.array(negative_data + positive_data)
    y = np.array([0] * len(negative_data) + [1] * len(positive_data))
   y = keras.utils.to_categorical(y)
   X_neg = X[:len(negative_data), ...]
   X_pos = X[len(negative_data):, ...]
   y_neg = y[:len(negative_data), ...]
   y_pos = y[len(negative_data):, ...]
    return X_neg, X_pos, y_neg, y_pos
def create_model():
   model = Sequential()
   model.add(Conv2D(CONV_FILTERS,
                     input_shape=(4, 100, 1),
                     kernel_size=KERNEL_SIZE,
                     activation="relu",
                     padding="same"))
   model.add(MaxPool2D(pool_size=POOL_SIZE))
    model.add(Flatten())
   model.add(Dense(HIDDEN_UNITS, activation="relu"))
   model.add(Dense(2, activation="softmax")) # same as 1 output sigmoid
    return model
MODEL_FUNC = create_model
```

```
def main():
   np.random.seed(1)
    TRAIN_TEST_FRAC = 0.9
    DATASET_SIZE = 5000
    # 10000 x (4, 100, 1) images total (5000 examples each)
    SPLIT = int(TRAIN_TEST_FRAC * DATASET_SIZE)
    Xn, Xp, yn, yp = get_x_y_data()
    shuffled_order = np.arange(0, DATASET_SIZE)
    np.random.shuffle(shuffled_order)
   Xn, Xp = Xn[shuffled_order, ...], Xp[shuffled_order, ...]
   yn, yp = yn[shuffled_order, ...], yp[shuffled_order, ...]
    X_train = np.vstack((Xn[:SPLIT, ...], Xp[:SPLIT, ...]))
    y_train = np.vstack((yn[:SPLIT, ...], yp[:SPLIT, ...]))
   X_test = np.vstack((Xn[SPLIT:, ...], Xp[SPLIT:, ...]))
    y_test = np.vstack((yn[SPLIT:, ...], yp[SPLIT:, ...]))
   print(X_train.shape)
    # define model
   model = MODEL_FUNC()
   model.compile(loss="categorical_crossentropy",
        optimizer="adam",
        metrics=["accuracy"])
    start = datetime.now()
    if MODEL_NAME == "lstm":
        X_train = X_train.squeeze()
        X_train = np.swapaxes(X_train, 1, 2)
        X_test = X_test.squeeze()
        X_test = np.swapaxes(X_test, 1, 2)
   model.fit(X_train, y_train, epochs=NUM_EPOCHS, batch_size=BATCH_SIZE)
    end = datetime.now()
    scores = model.evaluate(X_test, y_test)
    print("\n{}: {:.2f}%".format(model.metrics_names[1], scores[1] * 100))
    print("elapsed: {}".format(str(end - start)))
if __name__ == "__main__":
```

```
parser = argparse.ArgumentParser()
parser.add_argument(
    "-b". "--batch_size".
    help="Number of examples per batch",
    type=int)
parser.add_argument(
    "-e", "--epochs",
    help="Number of epochs to train over",
    type=int)
parser.add_argument(
    "-k", "--kernel",
    help="height, width tuple representing size of kernel.",
    type=str)
parser.add_argument(
    "-p", "--pool",
    help="height, width tuple representing size of pool.",
    type=str)
parser.add_argument(
    "-u", "--hidden_units",
    help="Number of hidden units for the Dense layer",
    type=int)
parser.add_argument(
    "-f", "--num_filters",
    help="Number of filters for the Conv layer",
    type=int)
parser.add_argument(
    "-m", "--model",
    help="Use a particular model implemented in alternative_models.py",
    type=str)
args = parser.parse_args()
if args.batch_size:
    BATCH_SIZE = args.batch_size
if args.epochs:
    NUM_EPOCHS = args.epochs
if args.kernel:
    height, width = map(int, args.kernel.split(","))
    KERNEL_SIZE = (height, width)
if args.pool:
   height, width = map(int, args.pool.split(","))
    POOL_SIZE = (height, width)
```

```
if args.hidden_units:
    HIDDEN_UNITS = args.hidden_units

if args.num_filters:
    CONV_FILTERS = args.num_filters

if args.model:
    MODEL_NAME = args.model
    MODEL_FUNC = getattr(models, args.model)

main()

# acc: 94.70%
# elapsed: 0:00:56.455801
```

(b) Layers

(c) Hyperparameters

Model	Hyperparameters	Train Accuracy	Test Accuracy	Training Time	
(1)	$epochs = 300$ $batch_size = 10$	100.00%	94.80%		
	$kernel_size = (4,4)$ $pool_size = (4,6)$ $hidden units = 32$			00:13:02	
	num_filters = 32				
(2)	$\begin{array}{c} \mathtt{epochs} = 50 \\ \mathtt{batch_size} = 10 \\ \mathtt{kernel_size} = (4,4) \\ \mathtt{pool_size} = (4,6) \\ \mathtt{hidden_units} = 32 \\ \mathtt{num_filters} = 1024 \\ \end{array}$	99.32%	95.60%	00:35:37	
(3)	$\begin{array}{c} \mathtt{epochs} = 50 \\ \mathtt{batch_size} = 10 \\ \mathtt{kernel_size} = (4,4) \\ \mathtt{pool_size} = (4,24) \\ \mathtt{hidden_units} = 512 \\ \mathtt{num_filters} = 32 \\ \end{array}$	97.56%	93.50%	00:02:34	

(d) Architectures

Model	Architecture	AC (Train)	AC (Test)	Time
deep_conv_net	Conv2D(32, (4, 6)) Conv2D(64, (4, 3)) MaxPool((4, 6)) Conv2D(128, (4, 3))	99.36%	98.10%	00:10:25
	Conv2D(1024, (1, 1)) Dense(64) Dense(2)			
${\tt fully_connected}$	$\begin{array}{c} \text{Dense}(1024) \\ \text{Dense}(256) \\ \text{Dense}(512) \\ \text{Dense}(2) \end{array}$	99.58%	93.20%	00:03:02
lstm	LSTM(64) Dense(2)	98.40%	98.30%	00:23:17