RBM

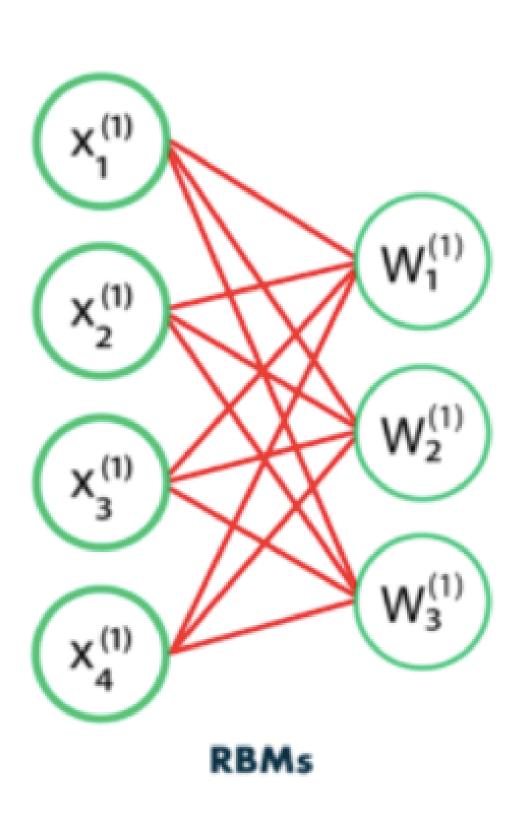
IMPLEMENTATION

AND TEST ON MNIST DATASET

ASSIGNMENT #3
BY-YOHANNIS KIFLE TELILA

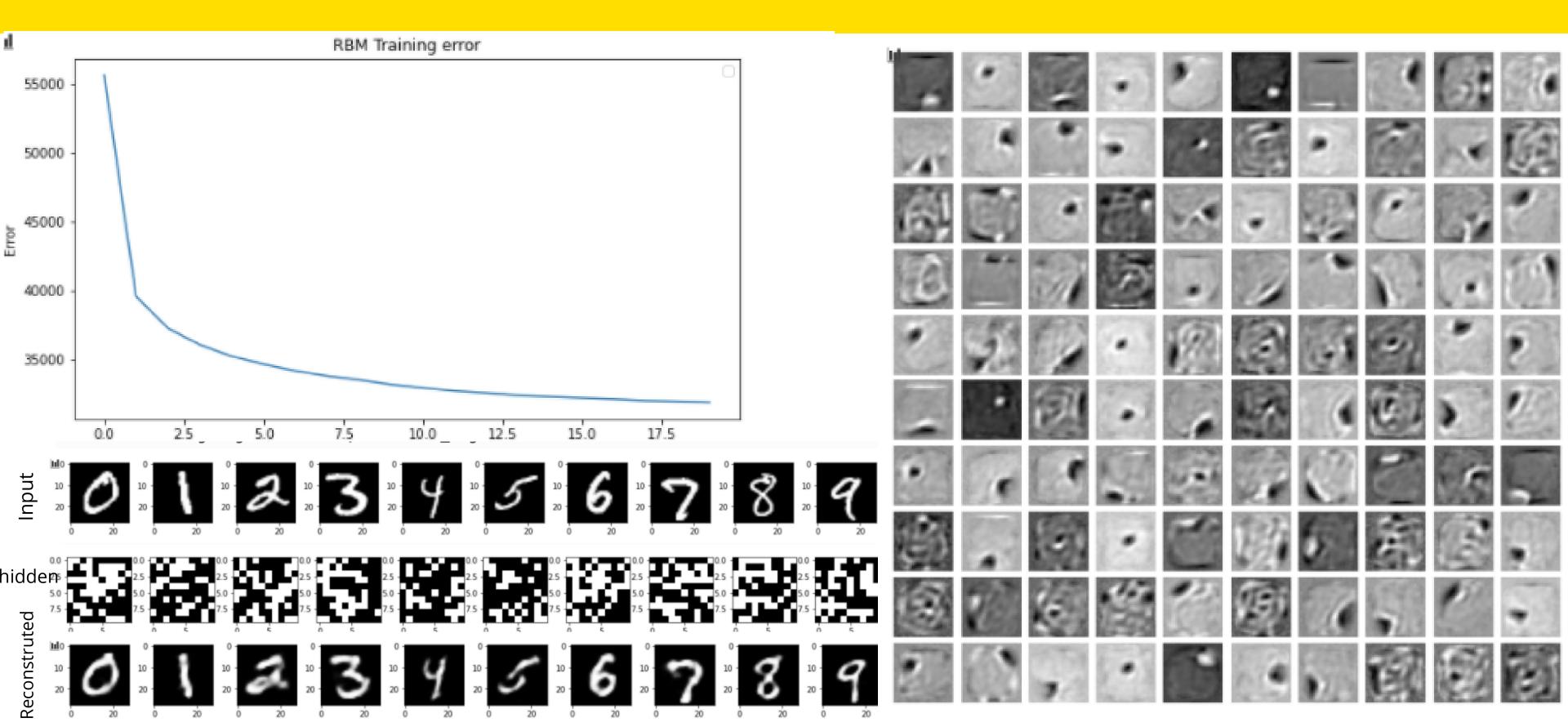


RBM MODEL



```
def posetivePhase(self, visibleLayer):
    # probability distribution of the hidden layer.
    pdH = expit(np.matmul(visibleLayer, self.vhW) + self.hlbias)
    return (pdH, np.random.binomial(1, p=pdH))
def negativePhase(self, hiddenLayer):
    # probability distribution of the visible layer.
    pdV = expit(np.matmul(hiddenLayer, self.vhW.T) + self.vlbias)
    return (pdV, np.random.binomial(1, p=pdV))
def compute_error_and_grads(self, batch):
    batchSize = batch.shape[0]
    v0 = batch.reshape(batchSize, -1)
   # Compute gradients - Positive Phase
   ph0, h0 = self.posetivePhase(v0)
   vhW_delta = np.matmul(v0.T, ph0)
   vb_delta = np.sum(v0, axis=0)
   hb_delta = np.sum(ph0, axis=0)
   # Compute gradients - Negative Phase
   pv1, v1 = self.negativePhase(h0)
   ph1, h1 = self.posetivePhase(pv1)
```

RBM RESULT



TESTING ON MNIST

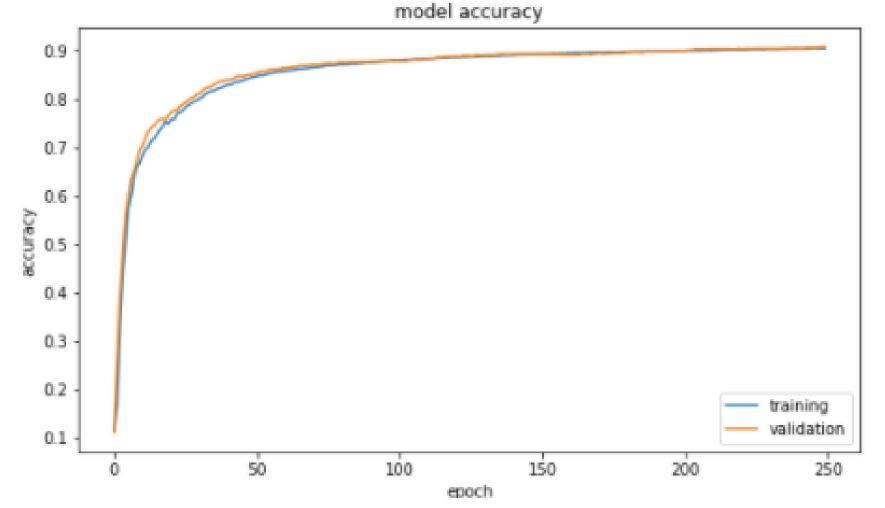
Encoding the representation

encoding = []

```
for i in range(len(trainingImages)):
    pdH, hiddenSampled = mnistRbm.posetivePhase(trainingImages[i].reshape(28*28))
    encoding.append(hiddenSampled)
encodedDfx = pd.DataFrame(encoding)
# svc model
                                                                          precision
                                                                                         recall f1-score
                                                                                                                support
model_linear = SVC(C=20, gamma=0.01, kernel="rbf")
                                                                                            0.97
                                                                                                        0.97
                                                                                                                    4721
                                                                                0.97
model_linear.fit(X_train, y_train)
                                                                                0.96
                                                                                            0.98
                                                                                                        0.97
                                                                                                                    5388
v pred = model linear.predict(X test)
                                                                                            0.95
                                                                                                        0.94
                                                                                0.93
                                                                                                                    4738
                                                                                0.93
                                                                                            0.91
                                                                                                        0.92
                                                                                                                    4926
accuracy: 0.9413541666666667
                                                                                0.95
                                                                                            0.93
                                                                                                        0.94
                                                                                                                    4681
                                                                                            0.92
                                                                                                        0.92
                                                                                0.92
                                                                                                                    4329
[[4598
                                            30
                                                                                0.96
                                                                                            0.97
                                                                                                        0.96
                                                                                                                    4758
    1 5307
              23
                                            12
                   10
                                                 11]
   16
                   34
                        19
        23 4524
                                            41
                                                 10]
                                                                                0.95
                                                                                            0.94
                                                                                                        0.94
                                                                                                                    5034
                            132
    17
         22
              80
                4504
                                                 36]
                                                                                0.93
                                                                                            0.90
                                                                                                        0.92
                                                                                                                    4665
              36
                                  34
    6
         20
                   12 4350
                                            20
                                                172]
                                                                      9
                                                                                0.91
                                                                                            0.92
                                                                                                        0.92
                                                                                                                    4760
              22
                  111
                        16 3976
                                            51
    34
        21
                                                 30]
              23
    37
         18
                        17
                             44 4595
                                                  0]
                                            16
                                                                                                        0.94
                                                                                                                   48000
              45
                                                              accuracy
    10
         30
                   37
                        40
                                                101]
                                     4750
                                            12
                                   0
    20
              61
                   90
                        27
                                                                                            0.94
                                                                                                        0.94
                                                                                                                   48000
        64
                             84
                                  38
                                       25 4205
                                                 51]
                                                                                0.94
                                                            macro avg
    11
              11
                       126
                                            38 4376]]
                                      110
                                                         weighted avg
                                                                                0.94
                                                                                            0.94
                                                                                                        0.94
                                                                                                                   48000
```

NN WITH SOFTMAX LAYER

```
image_size = 100 # 10*10
num_classes = 10 # ten unique digits
model = Sequential()
model.add(Dense(units=50, activation='sigmoid',kernel_initializer='uniform', input_shape=
  (image_size,)))
model.add(Dense(units=num_classes, activation='softmax'))
model.summary()
```



```
print(f'Test loss: {loss:.3}')
print(f'Test accuracy: {accuracy:.3}')

Test loss: 0.15
Test accuracy: 0.901
```

END

THANK YOU

github.com/joekifle