## Object Recognition and Image Understanding Exercise Sheet 7

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Task 1 see knn.py
Code:
        #!/usr/bin/env python2
        from prepare_toy_data import prepare_toy_data
        import random
        import math
        import operator
        import numpy as np
        from collections import Counter
        def train (X_train, y_train):
        # do nothing
        return
        def predict(X_train, y_train, x_test, k):
        # create list for distances and targets
        distances = []
        targets = []
        for i in range(len(X_train)):
        subtr = x_test - X_train
        distance = np.sqrt(np.sum(subtr))
        distances.append([distance, i])
        distances = sorted (distances)
        # make list of the k neighbours' targets
        for i in range(k):
        index = distances[i][1]
        targets.append(y_train[index])
```

```
return Counter(targets).most_common(1)[0][0]

def kNearestNeighbour(X_train, y_train, X_test, predictions, k):
# train on the input data
train(X_train, y_train)

for i in range(len(X_test)):
    predictions.append(predict(X_train, y_train, X_test[i, :], k))

def main():
    predictions = []
    X_train, X_test, labels_train, labels_test = prepare_toy_data(300, False k = 2
    kNearestNeighbour(X_train, labels_train, X_test, predictions, k)

predictions = np.asarray(predictions)

accuracy = accuracy_score(labels_test, predictions)
    print('Accuracy of this classifier: ' + accuracy*100 + '%')

main()
```

## Task 2

1. 
$$\mathcal{L}_{CE}(x) = -\log(p_{y}(x))$$

$$p_{y}(x) = p(class = y|x) = \frac{e^{sy(x)}}{\sum_{i=1}^{C} e^{s_{i}(x)}}$$

$$s_{i}(x) = \omega_{t}^{T} x + b_{i}$$

$$\frac{\partial s_{i}}{\partial \omega_{i}} = x$$

$$\frac{\partial s_{i}}{\partial s_{i}} = 1$$

$$\frac{\partial p_{y}}{\partial s_{i}} = p_{i}(1 - p_{i}), i = j$$

$$\frac{\partial p_{y}}{\partial s_{i}} = -p_{i}p_{j}, i \neq j$$

$$\frac{\partial \mathcal{L}_{CE}}{\partial p_{y}} = -\frac{1}{p_{y}} \frac{\partial p_{y}}{\partial s_{i}}$$

$$\frac{\partial \mathcal{L}_{CE}}{\partial \omega_{i}} = -\frac{1}{p_{y}} \frac{\partial p_{y}}{\partial s_{i}} p_{i}(1 - p_{i})x = (p_{i} - 1)x \text{ if } i = y$$

$$\frac{\partial \mathcal{L}_{CE}}{\partial \omega_{i}} = -\frac{1}{p_{y}} \frac{\partial p_{y}}{\partial s_{i}} - p_{i}p_{j}x = p_{i}x \text{ if } i \neq y$$

$$\frac{\partial \mathcal{L}_{CE}}{\partial b_{i}} = -\frac{1}{p_{y}} \frac{\partial p_{y}}{\partial s_{i}} p_{i}(1 - p_{i}) = (p_{i} - 1) \text{ if } i = y$$

$$\frac{\partial \mathcal{L}_{CE}}{\partial b_{i}} = -\frac{1}{p_{y}} \frac{\partial p_{y}}{\partial s_{i}} - p_{i}p_{j} = p_{i} \text{ if } i \neq y$$

2. See lin.py:

import numpy as np

```
from prepare_toy_data import prepare_toy_data
      from pr import prcurve
      class Perceptron(object):
      def = init_{-}(self, eta=0.01, n_iter=10):
               self.eta = eta
               self.n_iter = n_iter
      def fit (self, X, y):
               self.w_{-} = np.zeros(1 + X.shape[1])
               self.errors_{-} = []
      for _ in range(self.n_iter):
              errors = 0
                       for xi, target in zip(X, y):
                                update = self.eta * (target - self.predict(xi))
                                self.w_{-}[1:] += update * xi
                                self.w_{-}[0] += update
                                errors += int(update != 0.0)
                       self.errors_.append(errors)
              return self
      def net_input(self, X):
              return np. dot(X, self.w_[1:]) + self.w_[0]
      def predict (self, X):
              return np. where (self.net_input(X) \geq 0.0, 1, 0)
      X_train, X_test, labels_train, labels_test = prepare_toy_data(300, False
      alpha = 0.5
      iterations = 20
      p = Perceptron (alpha, iterations)
      p. fit (np. array (X_train), labels_train)
      predicted_test = p.predict(X_test)
      prcurve(predicted_test, labels_test)
3. See pr.py:
```

import numpy as np

```
import matplotlib.pyplot as plt
def prcurve (predicted_label, true_label):
TP = np. zeros (len (predicted_label))
FP = np.zeros(len(predicted_label))
FN = np.zeros(len(predicted_label))
for i in range(len(predicted_label)):
        if (predicted_label[i]==true_label[i]):
                TP[i]=1
        elif (predicted_label[i]==1 and true_label[i]==0):
                FP[i]=1
        elif(predicted\_label[i]==0 and true\_label[i]==1):
                FN[i]=1
recall=TP/(TP+FP)
precision=TP/(TP+FN)
precision2=precision.copy()
i=recall.shape[0]-2
# interpolation
while i \ge 0:
        if precision [i+1]>precision [i]:
                 precision[i] = precision[i+1]
        i=i-1
# plotting
fig, ax = plt.subplots()
for i in range (recall.shape [0]-1):
ax.plot((recall[i], recall[i]),(precision[i],
 precision[i+1]),'k-',label='',color='red') #vertical
ax. plot ((recall[i], recall[i+1]), (precision[i+1],
 precision[i+1]),'k-',label='',color='red') #horizontal
ax.plot(recall, precision2, 'k--', color='blue')
#ax.legend()
ax.set_xlabel("recall")
ax.set_ylabel("precision")
plt.savefig('fig.jpg')
fig.show()
```