

# ELE520

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## 1 Laboratory exercise 3

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### 1.1 Problem 1

The tasks are restructured to create a cleaner code and are mentioned throughout the notebook.

Importing numpy, creating the priors, and creating the matrices  $X_1$  and  $X_2$

```
[1]: import numpy as np
      from pdffuns import norm2D, knn2D, classplot
```

```
[2]: import pickle

pfile = 'lab3.p'
with open(pfile, 'rb') as fp:
    X = pickle.load(fp)
```

```
[3]: numX1 = len(X[0][0])
      numX2 = len(X[1][0])
      N = [numX1, numX2]
      numTot = numX1 + numX2
```

```
[4]: Pw = np.array([numX1/numTot, numX2/numTot])

      M = len(X)
```

```
[5]: x1 = np.linspace(start=-5, stop=10, num=25).reshape(-1, 1)
      x2 = np.linspace(start=-5, stop=10, num=25).reshape(-1, 1)

      X1, X2 = np.meshgrid(x1, x2)
```

Creating the main function to generate  $p(x)$ .

```
[6]: def labsol3(type='ML', title='', prm=1):
      pxw = []
      px = 0
```

```

if type == 'ML':
    # Subtask a) and b)
    my = []
    Sgm = []
    for dataset in X:
        my.append(np.mean(dataset, 1))
        Sgm.append(np.cov(dataset))

    for i in range(M):
        pxw.append(norm2D(my[i], Sgm[i], X1, X2))
        px += Pw[i] * pxw[i]
    overlapping = [True, False]
elif type == 'PZ':
    for i in range(M):
        h1 = prm

        hn = h1 / np.sqrt(N[i])
        hnI = hn**2 * np.eye(M)

        pns = []
        for j in range(N[i]):
            pns.append(norm2D(X[i][:, j], hnI, X1, X2))
        pxw.append(1/N[i] * np.sum(pns, axis=0))
        px += Pw[i] * pxw[i]
    overlapping = [True, False]
elif type == 'KN':
    for i in range(M):
        knn = prm

        pxw.append(knn2D(X[i], knn, X1, X2))
        px += Pw[i] * pxw[i]
    overlapping = [True, True]

g1 = []
g2 = []
for j in range(M):
    g1.append(Pw[j] * pxw[j]) # scaled class-conditional
    g2.append((pxw[j] * Pw[j]) / px) # posterior

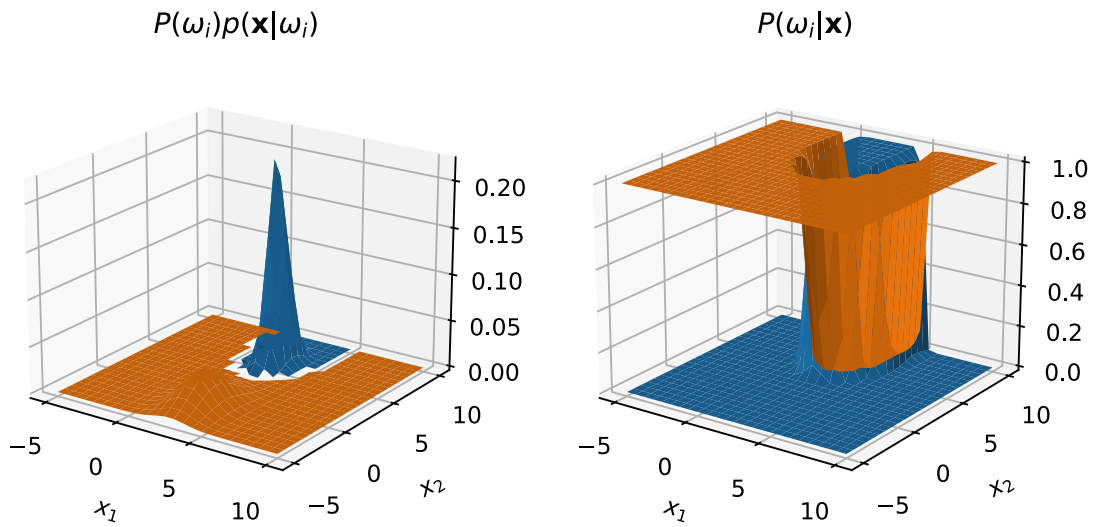
classplot([g1, g2], X1, X2, overlapping, title)

```

c) + d) for the maximum likelihood classification.

```
[7]: labso13('ML', 'Maximum Likelihood')
```

### Maximum Likelihood



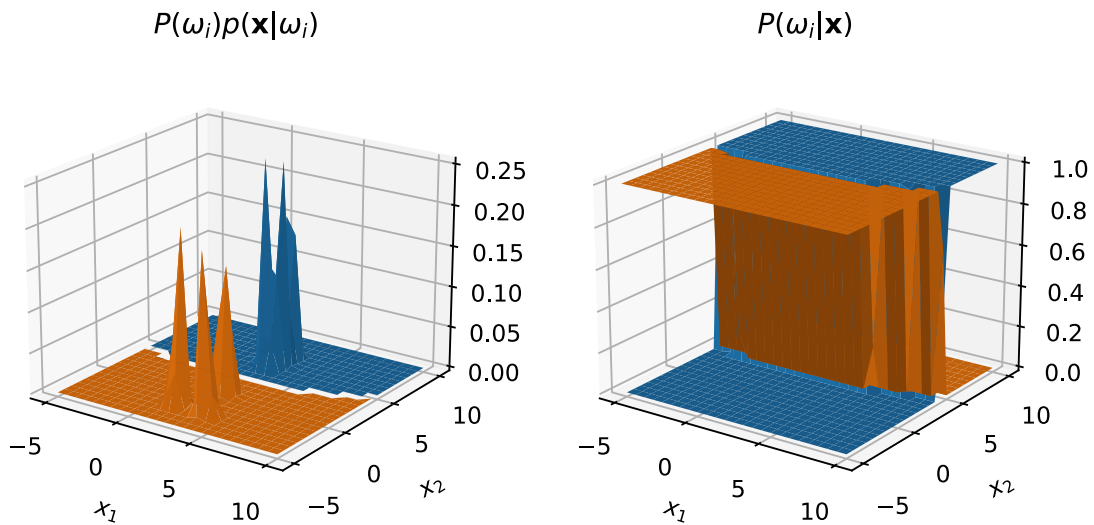
e) The decision boundary is marked by a tiny white line separating the two colors, and the decision areas are indicated by the different colors.

*Red* =  $\omega_1$  and *blue* =  $\omega_2$ .

c) + d) for the Parzen classifier with window size  $h_1 = 0.5$ .

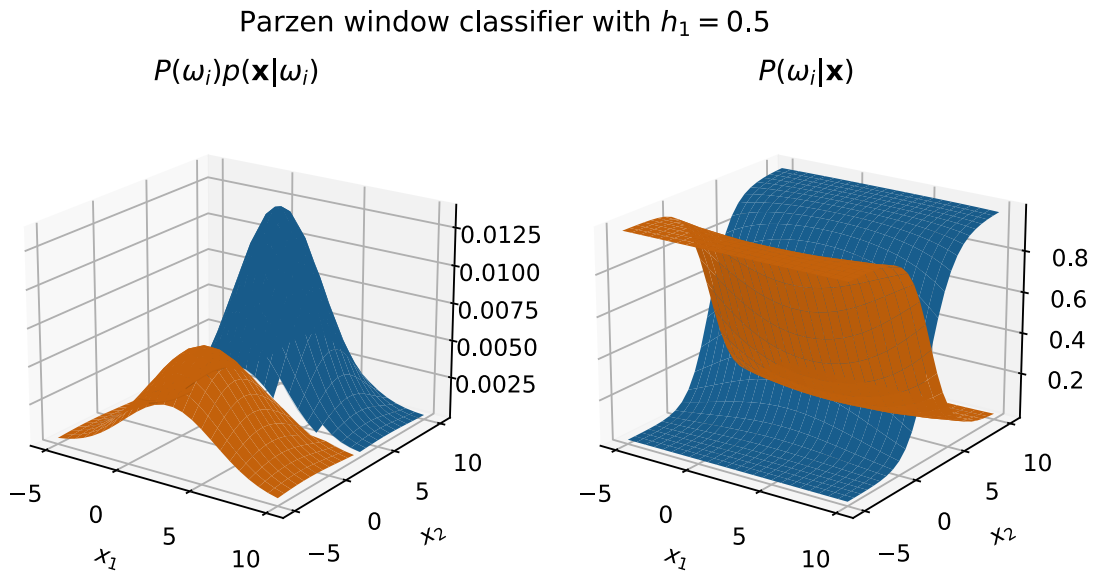
```
[8]: labsol3('PZ', ('Parzen window classifier with ' + r'$h_1 = 0.5$'), 0.5)
```

### Parzen window classifier with $h_1 = 0.5$



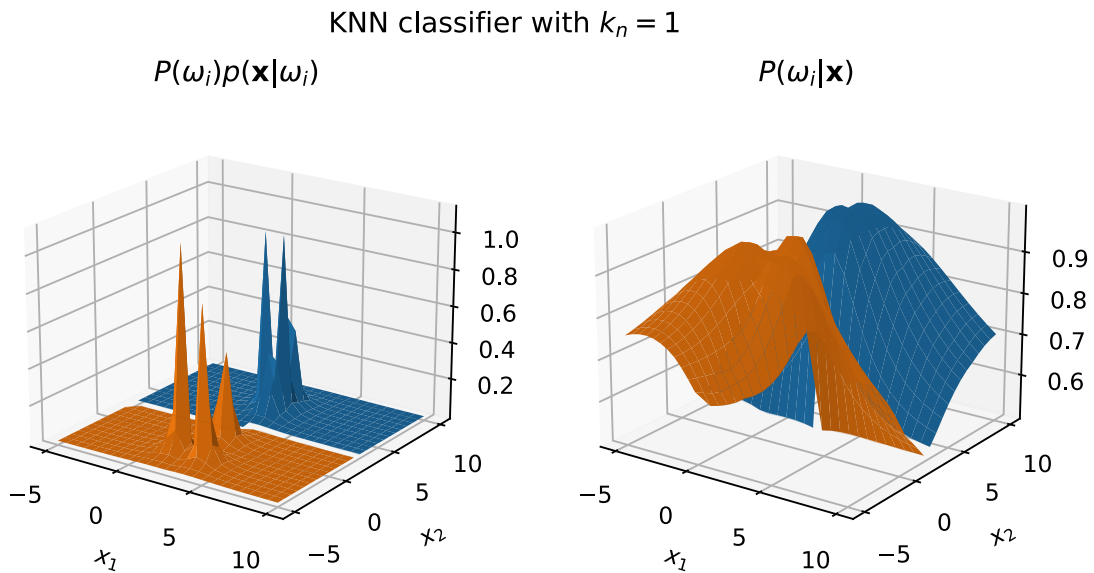
c) + d) for the Parzen classifier with window size  $h_1 = 5.0$ .

```
[9]: labso13('PZ', ('Parzen window classifier with ' + r'$h_1 = 0.5$'), 5.0)
```



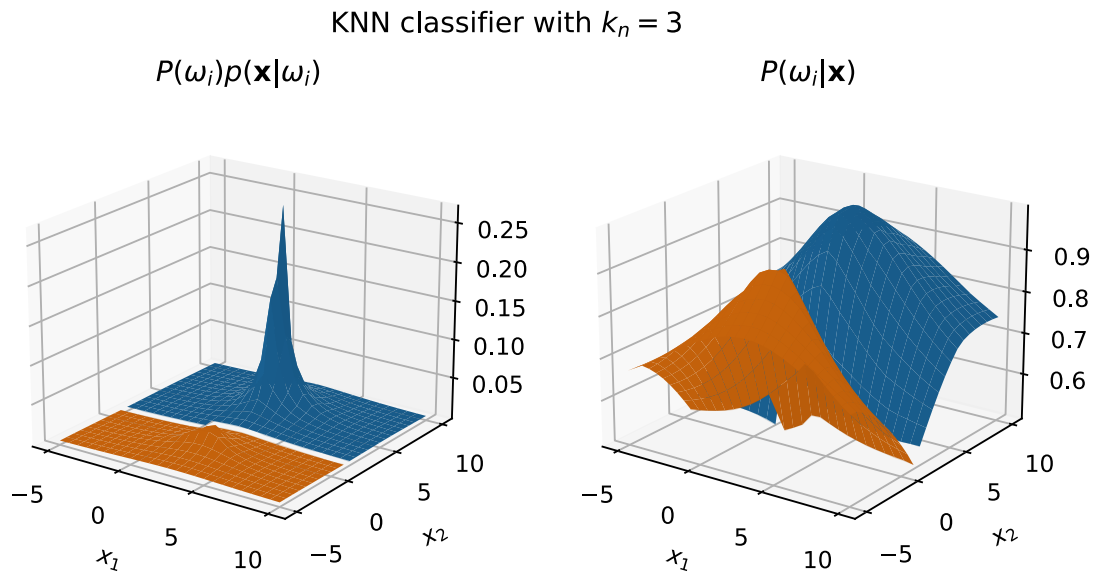
c) + d) for the knn-classifier with  $k_n = 1$ .

```
[10]: labso13('KN', ('KNN classifier with ' + r'$k_n = 1$'), 1)
```



c) + d) for the knn-classifier with  $k_n = 3$ .

```
[11]: labsol3('KN', ('KNN classifier with ' + r'$k_n = 3$'), 3)
```



c) + d) for the knn-classifier with  $k_n = 5$ . Calculating the knn-classifier with  $k_n = 5$  would not work as there is not enough data points. Because we only have 4 and 3 datapoints in each data set  $\chi_i$ , finding the 5th nearest neighbour is not possible.

The following is the content of pdfuns.py

```
[12]: import numpy as np
import matplotlib.pyplot as plt

def norm1D(my, Sgm, x):
    p = np.zeros(np.shape(x))
    n, _ = np.shape(x)

    const = 1 / (np.sqrt(2 * np.pi) * Sgm)

    for i in range(n):
        p[i] = const * \
            np.exp(-1 / 2 * np.square((x[i] - my)) / (np.square(Sgm)))

    return p

def norm2D(my, sigma, X1, X2):
    p = np.zeros(np.shape(X1))
    dim1, dim2 = np.shape(X1)
```

```

const = (2*np.pi)**(-len(my)/2) * (np.linalg.det(sigma))**(-1/2)

for i in range(dim1):
    for j in range(dim2):
        x_mu = ([X1[i, j], X2[i, j]] - my)
        p[i, j] = const * \
            np.exp(-1/2 * np.linalg.multi_dot([
                x_mu.T,
                np.linalg.inv(sigma),
                x_mu
            ]))
    return p

def knn2D(data, kn, X1, X2):
    p = np.zeros(np.shape(X1))
    dim1, dim2 = np.shape(X1)

    n = len(data[0])
    distances = []

    for i in range(n):
        my = data[:, i]
        distances.append(np.sqrt((X1 - my[0])**2 + (X2 - my[1])**2))

    for i in range(dim1):
        for j in range(dim2):
            distpoints = [distance[i][j] for distance in distances]
            idx = np.argsort(distpoints, 0)
            r = distpoints[idx[kn-1]]
            V = np.pi * r**2
            p[i, j] = kn / (n*V)

    return p

def classplot(every_g, X1, X2, overlappings=[True, False], title=''):
    fig = plt.figure(figsize=plt.figaspect(1/2))
    fig.suptitle(title)
    titles = [r'$P(\omega_i)p(\mathbf{x}|\omega_i)$',
              r'$P(\omega_i | \mathbf{x})$']

    for i, g in enumerate(every_g, 1):
        ax = fig.add_subplot(1, 2, i, projection='3d')
        ax.elev = 20
        ax.azim = -55

```

```

ax.set_title(titles[i-1])

for j in range(len(g)):
    new_g = np.copy(g[j])
    if overlappings[i-1]:
        overlap_mask = (new_g >= np.max(g, axis=0)).astype(float)
        np.putmask(new_g, overlap_mask == 0, np.nan)

    ax.plot_surface(X1, X2, new_g, facecolor=['r', 'b'])
ax.set(xlabel=r'$x_1$', ylabel=r'$x_2$')

plt.show()

```