

Exercises 3

1. A one-dimensional classification problem for three classes, A , B and C and pdf-s $f_A(x)$, $f_B(x)$, $f_C(x)$ is defined as:

$$p_A = 1/2, f_A(x) = 1/8 \text{ for } x \in [-4, 4]$$

$$p_B = 1/3, f_B(x) = 3(1 - x^2)/4 \text{ for } x \in [-1, 1]$$

$$p_C = 1/6, f_C(x) = x/8 \text{ for } x \in [0, 4]$$

Estimate optimal classification boundaries and decision rules for the system. Explain the general estimation steps and draw a figure that describes the estimation process and the results.

Answer: In general in a classification problem with a number of classes C_k , with *a priori* probabilities p_{C_k} and pdf's $f_{C_k}(x)$, the optimal decision rule for a given outcome x is to choose the class C_{opt} that maximizes the expression:

$$C_{opt} = \arg \max_{C_k} p_{C_k} \cdot f_{C_k}(x)$$

Here we set:

$$g_A(x) = p_A \cdot f_A(x) = \frac{1}{2} \cdot \frac{1}{8} = \frac{1}{16} \text{ when } x \in [-4, 4]$$

$$g_B(x) = p_B \cdot f_B(x) = \frac{1}{3} \cdot \frac{3(1-x^2)}{4} = \frac{1-x^2}{4} \text{ when } x \in [-1, 1]$$

$$g_C(x) = p_C \cdot f_C(x) = \frac{1}{6} \cdot \frac{x}{8} = \frac{x}{48} \text{ when } x \in [0, 4]$$

In order to find the classification borders we need to solve the following system of inequalities:

$$\begin{cases} g_A(x) > g_B(x) & (1) \\ g_A(x) > g_C(x) & (2) \\ g_B(x) > g_C(x) & (3) \end{cases}$$

so that $x \in C_k$, where

$$C_k = \begin{cases} A & \text{if } (1) \wedge (2) \\ B & \text{if } (3) \wedge \neg(1) \\ C & \text{if } \neg(2) \wedge \neg(3) \end{cases}$$

Now, let us solve this system. Equation (1) yields:

$$\frac{1}{16} > \frac{1-x^2}{4} \Leftrightarrow x^2 > \frac{3}{4} \Rightarrow \left\{ x > \frac{\sqrt{3}}{2} \vee x < -\frac{\sqrt{3}}{2} \right.$$

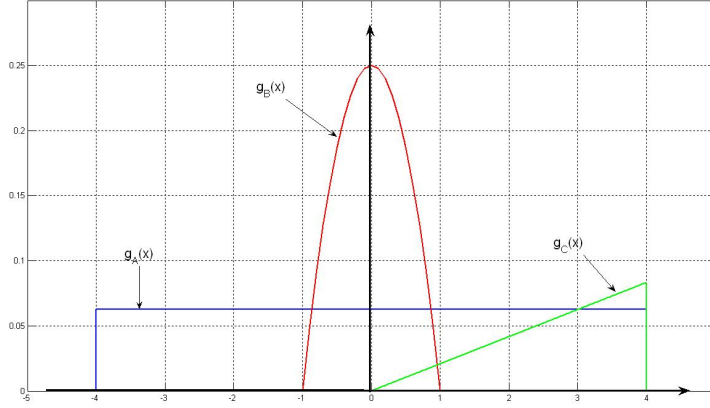


Figure 1: Illustration of the classification problem

Likewise, equation (2) yields:

$$\frac{1}{16} > \frac{x}{48} \Leftrightarrow x < 3$$

Before we calculate the solution to equation (3), let us plot the three functions $g_{C_k}(x)$. See Figure ??

We see that the solution to equation (3) is obsolete, since it's equivalent with the negated solution to equation (1) and the solution to equation (2). (Reason is that $3 > \sqrt{3}/2$). In other words, the classification boundaries are:

$$C_k = \begin{cases} A & \text{if } -4 < x < -\frac{\sqrt{3}}{2} \quad \vee \quad \frac{\sqrt{3}}{2} < x < 3 \\ B & \text{if } -\frac{\sqrt{3}}{2} < x < \frac{\sqrt{3}}{2} \\ C & \text{if } 3 < x < 4 \end{cases}$$

2. For a classification problem with two classes C_A and C_B , are the a priori probabilities $p_A = 3/4$ and $p_B = 1/4$. Assume the following pdf-s

$$p(\bar{z}|C_k) = \frac{1}{2\pi|\det\Sigma_k|^{1/2}} e^{-(\bar{z}-m_k)^T \Sigma_k^{-1} (\bar{z}-m_k)/2}$$

with

$$m_A = m_B = 0,$$

and

$$\Sigma_A = \begin{pmatrix} 4 & 0 \\ 0 & 1 \end{pmatrix}, \Sigma_B = \begin{pmatrix} 1 & 0 \\ 0 & 4 \end{pmatrix}.$$

Estimate decision boundaries for the problem!

Answer: The decision boundary is given by the "equal probability solution": $Pr(A|\bar{x}) = Pr(B|\bar{x})$. Now multiply each side with $Pr(\bar{x})$, giving $Pr(\bar{x}) \cdot Pr(A|\bar{x}) = Pr(\bar{x}) \cdot Pr(B|\bar{x})$. However, according to Bayes' rule, $Pr(\bar{x}) \cdot Pr(A|\bar{x}) = Pr(A) \cdot Pr(\bar{x}|A)$ which gives us the decision boundary is given by $Pr(A) \cdot Pr(\bar{x}|A) = Pr(B) \cdot Pr(\bar{x}|B)$. Thus we need to solve the equation

$$p_A \cdot p(\bar{x}|A) = p_B \cdot p(\bar{x}|B)$$

for $\bar{x} = [xy]$. Since we have the expression for $p(\bar{x}|C_k)$ and p_{C_k} , this is trivial. We will solve the equation as an inequality for the case $\bar{x} \in A$, i.e. $p_A \cdot p(\bar{x}|A) > p_B \cdot p(\bar{x}|B)$ (this is to have correct boundary conditions):

$$\begin{aligned} \frac{p_A}{2\pi |\det \Sigma_A|^{1/2}} e^{-\frac{(\bar{x}-m_A)^T \Sigma_A^{-1} (\bar{x}-m_A)}{2}} &> \frac{p_B}{2\pi |\det \Sigma_B|^{1/2}} e^{-\frac{(\bar{x}-m_B)^T \Sigma_B^{-1} (\bar{x}-m_B)}{2}} \Leftrightarrow \\ \frac{3}{4} \frac{1}{2\pi\sqrt{4}} e^{-\frac{[x \ y] \begin{bmatrix} 1/4 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}}{2}} &> \frac{1}{4} \frac{1}{2\pi\sqrt{4}} e^{-\frac{[x \ y] \begin{bmatrix} 1 & 0 \\ 0 & 1/4 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}}{2}} \Leftrightarrow \\ \ln 3 - [x/8 \ y/2] \begin{bmatrix} x \\ y \end{bmatrix} &> -[x/2 \ y/8] \begin{bmatrix} x \\ y \end{bmatrix} \Leftrightarrow \\ \ln 3 - \left(\frac{x^2}{8} + \frac{y^2}{2}\right) &> -\left(\frac{x^2}{2} + \frac{y^2}{8}\right) \Leftrightarrow \\ x^2 - y^2 + \frac{8}{3} \ln 3 &> 0 \end{aligned}$$

Now define $d(\bar{x}) \equiv d(x, y) \equiv x^2 - y^2 + \frac{8}{3} \ln 3$, then our classification rule becomes:

$$C_k = \begin{cases} A & \text{if } d(\bar{x}) > 0 \\ B & \text{if } d(\bar{x}) < 0 \end{cases}$$

Note that $d(x, y) = 0$ defines a hyperbolic curve, see Figure ??

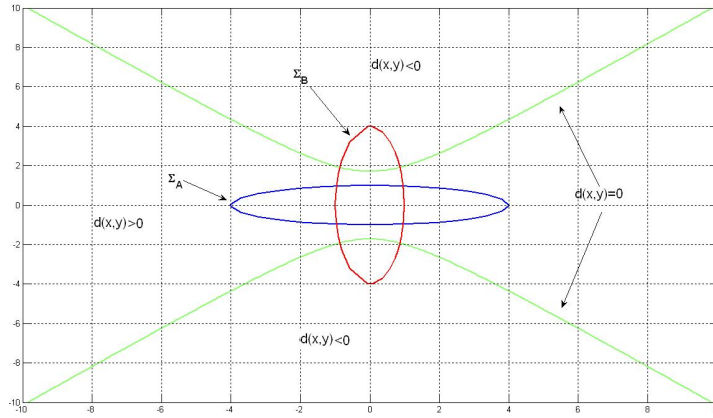
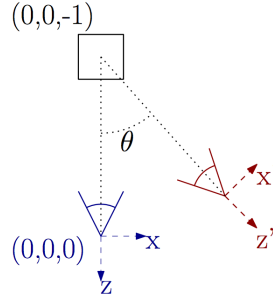


Figure 2: Illustration of the 2D classification problem

3. A robot is trying to gather 3D information from an object. Since it has only a single camera, it rotates around the object to obtain multiple views from it, as seen in the figure below.



- What is the relation between points P in 3D space and their image projections p on the image camera at $(0,0)$? Consider unit focal length, $f = 1$, and centered image origin, with x increasing to the right and y increasing up.
- What is the relation between 3D points $P = (x, y, z)$ in the original coordinate frame and $P' = (x', y', z')$ in the new coordinate frame after rotating θ radians?
- What is the relation between image points p and p' ?
- At some point the encoders of the motors fail and the robot doesn't know how large θ is. Estimate θ given point correspondences between the two images. How many point do you need?

Answer:

- Since z-axis points backwards, z-values are inverted in the projection.

$$P = (x, y, z) \Rightarrow p = \left(\frac{x}{-z}, \frac{y}{-z} \right)$$

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$$t = \begin{pmatrix} 0 \\ 0 \\ -1 \end{pmatrix}, \quad R = \begin{pmatrix} \cos \theta & 0 & -\sin \theta \\ 0 & 1 & 0 \\ \sin \theta & 0 & \cos \theta \end{pmatrix}$$

Transforming the 3D points look the same as transforming the camera, but in opposite direction. We first move the camera forward towards the object ($-t$), then rotate (R) and finally move back backwards (t).

$$P' = R(P - t) + t = R(P - (t - R^{-1}t)) = R(P - e)$$

Here e corresponds to the final translation of the camera.

$$e = t - R^{-1}t = \begin{pmatrix} 0 \\ 0 \\ -1 \end{pmatrix} - \begin{pmatrix} \cos \theta & 0 & \sin \theta \\ 0 & 1 & 0 \\ -\sin \theta & 0 & \cos \theta \end{pmatrix} \begin{pmatrix} 0 \\ 0 \\ -1 \end{pmatrix} = \begin{pmatrix} \sin \theta \\ 0 \\ -1 + \cos \theta \end{pmatrix}$$

- There is an essential matrix $E = RT_e$ for which $P'^T EP = 0$ holds.

$$T_e = \begin{pmatrix} 0 & 1 - \cos \theta & 0 \\ -(1 - \cos \theta) & 0 & -\sin \theta \\ 0 & \sin \theta & 0 \end{pmatrix}$$

$$E = \begin{pmatrix} \cos \theta & 0 & -\sin \theta \\ 0 & 1 & 0 \\ \sin \theta & 0 & \cos \theta \end{pmatrix} \begin{pmatrix} 0 & 1 - \cos \theta & 0 \\ -(1 - \cos \theta) & 0 & -\sin \theta \\ 0 & \sin \theta & 0 \end{pmatrix}$$

$$E = \begin{pmatrix} 0 & -(1 - \cos \theta) & 0 \\ -(1 - \cos \theta) & 0 & -\sin \theta \\ 0 & \sin \theta & 0 \end{pmatrix}$$

Then the projections can be related by

$$\begin{pmatrix} p'_x & p'_y & -1 \end{pmatrix} E \begin{pmatrix} p_x \\ p_y \\ -1 \end{pmatrix} = 0$$

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$$\begin{pmatrix} p'_x & p'_y & -1 \end{pmatrix} \begin{pmatrix} 0 & -(1 - \cos \theta) & 0 \\ -(1 - \cos \theta) & 0 & -\sin \theta \\ 0 & \sin \theta & 0 \end{pmatrix} \begin{pmatrix} p_x \\ p_y \\ -1 \end{pmatrix} = 0$$

$$\begin{pmatrix} p'_x & p'_y & -1 \end{pmatrix} \begin{pmatrix} -(1 - \cos \theta)p_y \\ -(1 - \cos \theta)p_x + \sin \theta \\ \sin \theta p_y \end{pmatrix} = 0$$

$$-(1 - \cos \theta)p_y p'_x - (1 - \cos \theta)p_x p'_y + \sin \theta p'_y - \sin \theta p_y = 0$$

$$(p_y p'_x + p_x p'_y) \cos \theta + (p'_y - p_y) \sin \theta = (p_y p'_x + p_x p'_y)$$

Finally, solve for θ by removing the cosine and then using arcsin.

4. Assume that you have a robot translating in space. For every point in time a camera (with focal length f), mounted on the robot, observes two points A and B by measuring the image coordinates (x_A, y_A) and (x_B, y_B) , as well as the motion field (x'_A, y'_A) and (x'_B, y'_B) . Set up a linear system of equations from which is possible to determine the translational direction of the robot (T_X, T_Y, T_Z) up to an unknown scale factor, including the depths of the two points Z_A and Z_B . Explain how it is possible to determine the direction, even if the scale factor is unknown.

Answer: By derivating the equation for the perspective mapping

$$\frac{x}{f} = \frac{X}{Z}$$

$$\frac{y}{f} = \frac{Y}{Z}$$

we get

$$\frac{\dot{x}}{f} = \frac{\dot{X}}{Z} - \frac{X}{Z} \frac{\dot{Z}}{Z}$$

$$\frac{\dot{y}}{f} = \frac{\dot{Y}}{Z} - \frac{Y}{Z} \frac{\dot{Z}}{Z}$$

With $(\dot{X}, \dot{Y}, \dot{Z})^T = -(T_X, T_Y, T_Z)^T$, corresponding to the a pure translation, you get

$$\begin{aligned}\frac{\dot{x}}{f} &= -\frac{T_X}{Z} + \frac{x}{f} \frac{T_Z}{Z} \\ \frac{\dot{y}}{f} &= -\frac{T_Y}{Z} + \frac{y}{f} \frac{T_Z}{Z}\end{aligned}$$

and if you multiply with fZ you get

$$\begin{aligned}\dot{x}Z + fT_X &= xT_Z \\ \dot{y}Z + fT_Y &= yT_Z\end{aligned}$$

Suppose we have the measurements (x, y) and (\dot{x}, \dot{y}) for two points A och B . We can then write the equations above as a system of equations

$$\begin{pmatrix} f & 0 & \dot{x}_A & 0 \\ 0 & f & \dot{y}_A & 0 \\ f & 0 & 0 & \dot{x}_B \\ 0 & f & 0 & \dot{y}_B \end{pmatrix} \begin{pmatrix} T_X \\ T_Y \\ Z_A \\ Z_B \end{pmatrix} = T_Z \begin{pmatrix} x_A \\ y_A \\ x_B \\ y_B \end{pmatrix}$$

to determine Z , T_X and T_Y as a function of T_Z , which is not known. This means that we can only determine the motion (T_X, T_Y, T_Z) and depths (Z_A and Z_B) up to some unknown scale factor. The translational direction though is determined by

$$\left(\frac{T_X}{T_Z}, \frac{T_Y}{T_Z}, 1 \right)^T,$$

which can be determined, but in order to know the speed we need more information, such as e.g. Z_A or Z_B .