

# Problem Set 1

19fmiROBEG - AI Robotics

## 1 Problems

**Problem 1.** (0.5 points) Given the two random variables  $x = [2, 4, 5, 1, 9, 11, 6]$  and  $y = [4, 1, 8, 2, 7, 10, 6]$  calculate the  $2 \times 2$  covariance matrix for these two random variables.

**Problem 2.** (0.5 points) In the lectures state estimation using a Kalman Filter was applied to the case of tracking the position and velocity of a vehicle in one-dimension was shown. Given a vehicle in 3D space, modelled with cartesian coordinates  $x, y, z$ . Determine the state vector  $x_k$ , control vector  $u_k$ , the transition matrix  $A$  and control-input matrix  $B$ .

**Problem 3.** (2 point) Starting from the covariance update equation

$$P_k = (I - K_k H_k) P_k^- (I - K_k H_k)^T + K_k R_k K_k^T$$

show that the Kalman Gain  $K = P_k^- H_k^T [H_k P_k^- H_k^T + R_k]^{-1}$

Note the superscript minus symbol in  $P_k^-$  indicates that this is the error covariance obtained in the first prediction stage

Hint:  $\frac{\partial Tr(P_k)}{\partial K_k} = 0$

**Problem 4.** (2 points) The Kalman filter is limited to linear models, however the extension to non-linear models is possible through the use of the Extended Kalman filter. This is achieved by linearising the model around the current state. We replace the state vector  $x$  with a new, possibly non-linear function  $g$ . Linearisation of the function  $g$  about the time-step  $k$  is given by computing the Taylor series expansion of  $g$  to first order.

In 2D the robot has a state vector  $x_k = [x_k, y_k, \theta_k]^T$  and control vector  $v_k = [V_{x,k}, V_{y,k}, \omega_k]^T$ . Its motion model  $g()$  can be described as the following

$$\begin{aligned}x_k &= x_{k-1} + \Delta t V_{x,k-1} \times \cos(\theta_{k-1}) \\y_k &= y_{k-1} + \Delta t V_{y,k-1} \times \sin(\theta_{k-1}) \\\theta_k &= \theta_{k-1} + \omega_{k-1}\end{aligned}$$

We can calculate the Jacobian of  $G$  to provide the first order differential of the motion model. Given as

$$G = \begin{pmatrix} \frac{\partial x_k}{\partial x} & \frac{\partial x_k}{\partial y} & \frac{\partial x_k}{\partial \theta} \\ \frac{\partial y_k}{\partial x} & \frac{\partial y_k}{\partial y} & \frac{\partial y_k}{\partial \theta} \\ \frac{\partial \theta_k}{\partial x} & \frac{\partial \theta_k}{\partial y} & \frac{\partial \theta_k}{\partial \theta} \end{pmatrix}$$

Calculate the Jacobian  $G$ .