

# Kalman Filtering: Problems

ELG 5218 – Uncertainty Evaluation in Engineering Measurements and Machine Learning

Instructor: Miodrag Bolić, University of Ottawa

## Contents

<b>1</b>	<b>Joint Distributions and Correlation</b>	<b>2</b>
<b>2</b>	<b>State Space Models and Dynamic Systems</b>	<b>2</b>
<b>3</b>	<b>Kalman Filter Fundamentals</b>	<b>2</b>

# 1 Joint Distributions and Correlation

## Problem 1.1: 2D Joint Distribution

Consider two random time series with means  $\mu_1 = 1$ ,  $\mu_2 = 2$ , variances  $\sigma_1^2 = 1$ ,  $\sigma_2^2 = 2$ , and correlation coefficient  $\rho = 0.5$ .

1. Write the covariance matrix  $\Sigma$ .
2. What is the covariance  $\text{COV}(z_1, z_2)$ ?
3. Describe the shape of the 2D Gaussian distribution (e.g., orientation, spread).
4. How would the shape change if  $\rho = -0.5$ ? If  $\rho = 0$ ?

## Problem 1.2: Covariance and Correlation

1. Define the sample covariance between two time series  $z_1(t)$  and  $z_2(t)$  with  $N$  samples.
2. Define the correlation coefficient  $\rho(z_1, z_2)$  and explain its range and interpretation.
3. If  $\text{COV}(z_1, z_2) = 3.0$ ,  $\sigma_1 = 2$ , and  $\sigma_2 = 3$ , compute the correlation coefficient.
4. What does  $\rho = 0.9$  imply about the relationship between  $z_1$  and  $z_2$ ?

## Problem 1.3: Multivariate Gaussian Transformation

Consider a multivariate Gaussian vector  $\mathbf{z} = [z_1, z_2]^\top$  with mean  $\boldsymbol{\mu}$  and covariance  $\Sigma_z$ .

Let  $\mathbf{y} = A\mathbf{z} + \mathbf{b}$  where  $A$  is a  $2 \times 2$  matrix.

1. Derive the mean of  $\mathbf{y}$ .
2. Derive the covariance of  $\mathbf{y}$ .
3. If  $A = \begin{bmatrix} 2 & 0.5 \\ 0 & 1 \end{bmatrix}$ ,  $\mathbf{b} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$ ,  $\boldsymbol{\mu} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$ , and  $\Sigma_z = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$ , compute  $\boldsymbol{\mu}_y$  and  $\Sigma_y$ .

# 2 State Space Models and Dynamic Systems

## Problem 2.1: Constant Velocity Model

An object moves in 2D with constant velocity. State:  $\mathbf{z}_t = [z_{1,t}, \dot{z}_{1,t}, z_{2,t}, \dot{z}_{2,t}]^\top$ .

1. Derive the state transition matrix  $F$  for sampling interval  $\Delta t$ .
2. If only position  $(z_1, z_2)$  is measured, what is  $H$ ?

# 3 Kalman Filter Fundamentals

## Problem 3.1: Prediction and Update Steps

Consider the scalar system:

$$z_{t+1} = z_t + q_t, \quad y_t = z_t + r_t$$

with  $q_t \sim \mathcal{N}(0, Q)$ ,  $r_t \sim \mathcal{N}(0, R)$ , and initial estimate  $\hat{z}_{0|0} = 5$ , true value  $z_0 = 10$ .

1. Describe the prediction step: what does the filter predict for  $\hat{z}_{1|0}$ ?
2. If measurement  $y_1 = 11$  is observed, describe the update step.
3. Write the update equation:  $\hat{z}_{1|1} = \hat{z}_{1|0} + K_1(y_1 - \hat{z}_{1|0})$ .
4. Explain what the Kalman gain  $K_1$  represents.

### Problem 3.2: Kalman Gain Properties

The Kalman gain is defined as:

$$K_t = \frac{\Sigma_{t|t-1}}{\Sigma_{t|t-1} + R}$$

1. What happens to  $K_t$  if  $R \rightarrow 0$  (very accurate measurements)?
2. What happens to  $K_t$  if  $R \rightarrow \infty$  (very noisy measurements)?
3. What happens to  $K_t$  if  $\Sigma_{t|t-1} \rightarrow 0$  (very confident prediction)?
4. Interpret the Kalman gain as a confidence-weighted average.

### Problem 3.3: Covariance Evolution

Starting with  $\Sigma_{0|0} = 100$  (very uncertain),  $Q = 0.01$ ,  $R = 4$ :

1. Compute  $\Sigma_{1|0}$  (predicted variance).
2. Compute  $K_1$ .
3. Compute  $\Sigma_{1|1}$  (updated variance).
4. Repeat for  $t = 2, 3, 4$  and describe the trend.
5. At what iteration does steady state occur?

### Problem 3.4: 2D constant-velocity model: matrix shapes & one update

State  $\mathbf{z}_t = [x_t, \dot{x}_t, y_t, \dot{y}_t]$  with  $\Delta t = 1$  and

$$\mathbf{F} = \begin{bmatrix} 1 & 1 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 \end{bmatrix}, \quad \mathbf{H} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix}.$$

Assume  $\mathbf{Q} = 0.1 \mathbf{I}_4$ ,  $\mathbf{R} = \text{diag}(1, 4)$ , prior  $\boldsymbol{\mu}_{t-1|t-1} = 0$ ,  $\boldsymbol{\Sigma}_{t-1|t-1} = \mathbf{I}_4$ , no control inputs, and measurement  $\mathbf{y}_t = [1, -2]$ .

1. Compute  $\boldsymbol{\mu}_{t|t-1}$  and  $\boldsymbol{\Sigma}_{t|t-1}$ .
2. Compute  $\hat{\mathbf{y}}_t$ ,  $\mathbf{S}_t$ , and  $\mathbf{K}_t$ .
3. Compute the updated mean  $\boldsymbol{\mu}_{t|t}$  (you may leave  $\boldsymbol{\Sigma}_{t|t}$  in symbolic form).
4. Intuition: which coordinate ( $x$  or  $y$ ) should be corrected more strongly, and why?