# Filtering and Smoothing Exercise

## Problem 1: Filtering and Smoothing

Given the state-space model:

$$X_{n+1} = \frac{1}{2}X_n + 1 + \Xi_n, \quad \Xi_n \sim \mathcal{N}(0, 1)$$
$$Y_n = X_n + \sqrt{2}\Sigma_n, \quad \Sigma_n \sim \mathcal{N}(0, 1)$$
$$X_0 \sim \mathcal{N}(-1, 2)$$

### 1. Prediction step: Distribution of $X_1$

The state transition is:

$$X_1 = \frac{1}{2}X_0 + 1 + \Xi_0$$

With:

$$\mathbb{E}[X_1] = \frac{1}{2}(-1) + 1 = \frac{1}{2}, \quad \text{Var}(X_1) = \left(\frac{1}{2}\right)^2 \cdot 2 + 1 = \frac{3}{2}$$

$$X_1 \sim \mathcal{N}\left(\frac{1}{2}, \frac{3}{2}\right)$$

# 2. Filtering step: Distribution of $X_1 \mid Y_1 = 2$

Using Kalman filter update:

$$K = \frac{\frac{3}{2}}{\frac{3}{2} + 2} = \frac{3}{7}, \quad \mathbb{E}[X_1 \mid Y_1 = 2] = \frac{1}{2} + \frac{3}{7} \left(2 - \frac{1}{2}\right) = \frac{8}{7}$$

$$\operatorname{Var}(X_1 \mid Y_1 = 2) = \left(1 - \frac{3}{7}\right) \cdot \frac{3}{2} = \frac{6}{7}$$

$$X_1 \mid Y_1 = 2 \sim \mathcal{N}\left(\frac{8}{7}, \frac{6}{7}\right)$$

#### 3. Smoothing step: Distribution of $X_0 \mid Y_1 = 2$

The observation model is:

$$Y_1 = \frac{1}{2}X_0 + 1 + \underbrace{\Xi_0 + \sqrt{2}\Sigma_1}_{N(0,3)}$$

Conditional distribution parameters:

$$\operatorname{Var}(X_0 \mid Y_1 = 2) = \left(\frac{1}{2} + \frac{(1/2)^2}{3}\right)^{-1} = \frac{12}{7}$$

$$\mathbb{E}[X_0 \mid Y_1 = 2] = \frac{12}{7} \left( \frac{-1}{2} + \frac{(1/2)(2-1)}{3} \right) = -\frac{4}{7}$$

$$X_0 \mid Y_1 = 2 \sim \mathcal{N}\left( -\frac{4}{7}, \frac{12}{7} \right)$$

#### 4. Pseudo Code for General Model

Model parameters:

$$X_{n+1} = \alpha X_n + \beta \Xi_n, \quad \Xi_n \sim \mathcal{N}(0, 1)$$
 
$$X_0 \sim \mathcal{N}(m, 1)$$
 
$$Y_n = X_n + \sqrt{2}\Sigma_n, \quad \Sigma_n \sim \mathcal{N}(0, 1)$$
 Input:  $\alpha, \beta, m, y \ (Y_1 = y)$ 

#### Algorithm 1 Compute Distributions

```
1: function COMPUTEDISTRIBUTIONS(\alpha, \beta, m, y)
          // Prediction: X_1
          pred\_mean \leftarrow \alpha \cdot m
 3:
          pred_var \leftarrow \alpha^2 \cdot 1 + \beta^2
 4:
                                                                                                                     \triangleright Note: initial variance = 1
          // Filtering: X_1 \mid Y_1 = y
 5:
          K \leftarrow \text{pred\_var}/(\text{pred\_var} + 2)
                                                                                                             \triangleright Observation noise variance = 2
 6:
          filter\_mean \leftarrow pred\_mean + K \cdot (y - pred\_mean)
 7:
          filter_var \leftarrow (1 - K) \cdot pred_var
 8:
          // Smoothing: X_0 \mid Y_1 = y
9:
          denom \leftarrow \alpha^2 \cdot 1 + \beta^2 + 2
                                                                                                                \triangleright \alpha^2 \text{Var}(X_0) + \beta^2 + \text{obs noise}
10:
11:
          K_s \leftarrow \alpha/\text{denom}
          smooth_mean \leftarrow m + K_s \cdot (y - \alpha \cdot m)
12:
                                                                                                                        \triangleright \operatorname{Var}(X_0) - \alpha K_s \operatorname{Var}(X_0)
          smooth_var \leftarrow 1 - \alpha \cdot K_s
13:
          return (pred_mean, pred_var), (filter_mean, filter_var), (smooth_mean, smooth_var)
14:
15: end function
```

#### Output:

- (i)  $X_1$ :  $\mathcal{N}(\text{pred\_mean, pred\_var})$
- (ii)  $X_1 \mid Y_1 = y$ :  $\mathcal{N}(\text{filter\_mean}, \text{filter\_var})$
- (iii)  $X_0 \mid Y_1 = y$ :  $\mathcal{N}(\text{smooth\_mean}, \text{smooth\_var})$

**Note:** This pseudo code assumes no constant term in the state transition equation. For models with constants (like the +1 in the original problem), additional adjustments would be needed.