

**Exercise 07 for MA-INF 2201 Computer Vision WS24/25**  
**Submission on 15.12.2024**

## Overview

This exercise focuses on state estimation and tracking, combining theoretical foundations with practical implementation. You will implement a Kalman filter and a fixed-lag smoother to track a moving object in 2D space. Use the numpy and matplotlib libraries for implementation and visualization. Provide detailed comments for your implementation.

1. **Kalman Filter Implementation:** Consider a 2D signal with acceleration. The state vector includes position, velocity, and acceleration components:  $\mathbf{x} = [x, y, v_x, v_y, a_x, a_y]^T$ .  
**Data:** The observation data is provided in 'data/observations.npy' file which contains noisy 2D position measurements file which can be loaded using:

```
observations = np.load('observations.npy') # Shape: (N, 2)
```

### System Model:

- Time step:  $\Delta t = 0.1$  seconds
- State transition matrix  $\Psi$  incorporating acceleration terms:

$$\Psi = \begin{bmatrix} 1 & 0 & \Delta t & 0 & \frac{\Delta t^2}{2} & 0 \\ 0 & 1 & 0 & \Delta t & 0 & \frac{\Delta t^2}{2} \\ 0 & 0 & 1 & 0 & \Delta t & 0 \\ 0 & 0 & 0 & 1 & 0 & \Delta t \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix}$$

- Process Noise Covariance ( $\Sigma_p$ ): Diagonal matrix with process noise parameter  $sp = 0.001$
- Measurement Matrix ( $\Phi$ ) observing only positions:

$$\Phi = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 \end{bmatrix}$$

- Measurement Noise Covariance ( $\Sigma_m$ ): Diagonal matrix with measurement noise parameter  $sm = 0.05$

- 1.1. What should be the time evolution equation? Explain each term. (1 point)
  - 1.2. What should be the measurement equation? Explain each term. (1 point)
  - 1.3. Implement a Kalman filter to track the signal: Initialize the state vector with  $\mathbf{x}_0 = [-10, -150, 1, -2, 0, 0]^T$ . Implement the prediction step (time update). and the correction step (measurement update). (6 points)
  - 1.4. Visualize the observations vs. filtered estimates. (2 points)
2. **Fixed-Lag Smoothing:** Extend your Kalman filter implementation to perform fixed-lag smoothing for improved state estimation.
    - 2.1. Implement the smoother with a lag of 5 time steps. (6 points)
    - 2.2. Analysis of lag size effects. (4 points)

## Submission Guidelines

- Submit your implementation in a Python file named `solution.py`, which should include two classes: `KalmanFilter` and `FixedLagSmoothing`.
- Include clear documentation and comments in your code.
- Provide theoretical answers and analysis (1.1, 1.2, explanation of 2.2) in `answers.pdf`. (Please limit your answers to no more than 15 lines.)