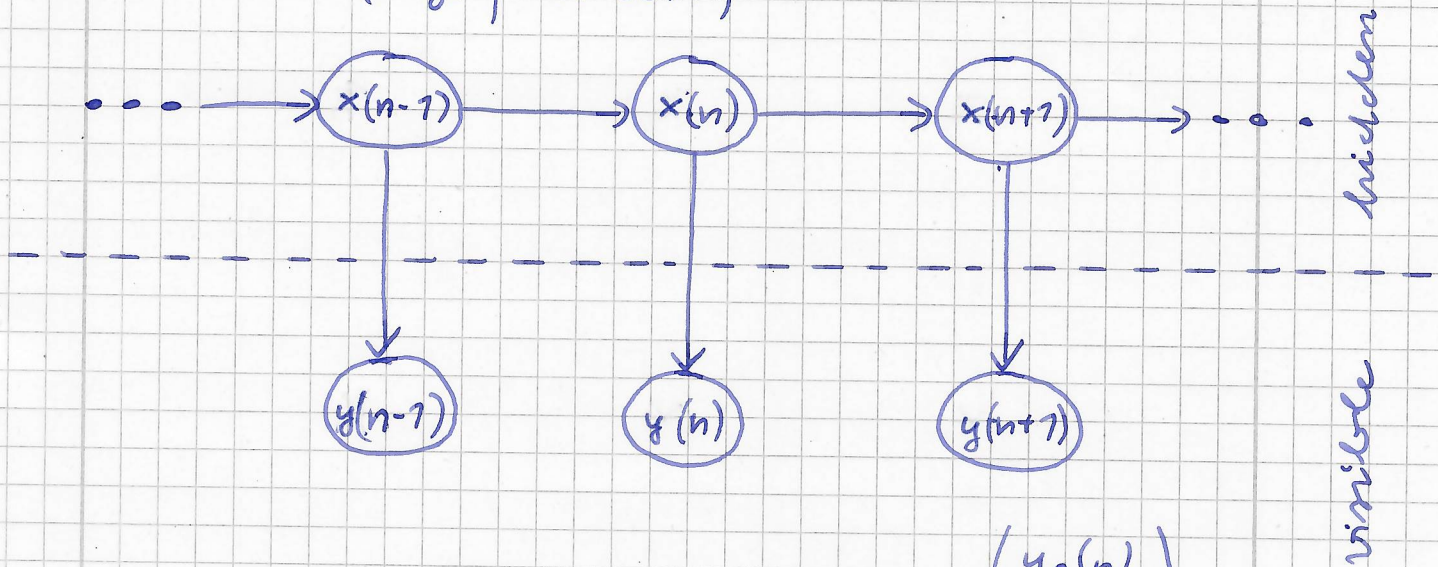


Dynamical Systems

Hidden Markov Model

$$\text{state vector } x(n) = \begin{pmatrix} x_1(n) \\ \vdots \\ x_k(n) \end{pmatrix}$$

(e.g. position)



$$\text{observation vector } y(n) = \begin{pmatrix} y_1(n) \\ \vdots \\ y_L(n) \end{pmatrix}$$

(measurements, e.g., ranges)

- discrete time index n
- arrows indicate statistical dependence

Linear Dynamical System Model

- for simplicity zero mean noise
- state transition equation

$$x(n+1) = F(n) \cdot x(n) + u(n)$$



zero mean state noise,
correlation matrix

$$R_{uu}(n) = E\{u(n) \cdot u^T(n)\}$$

- observation equation

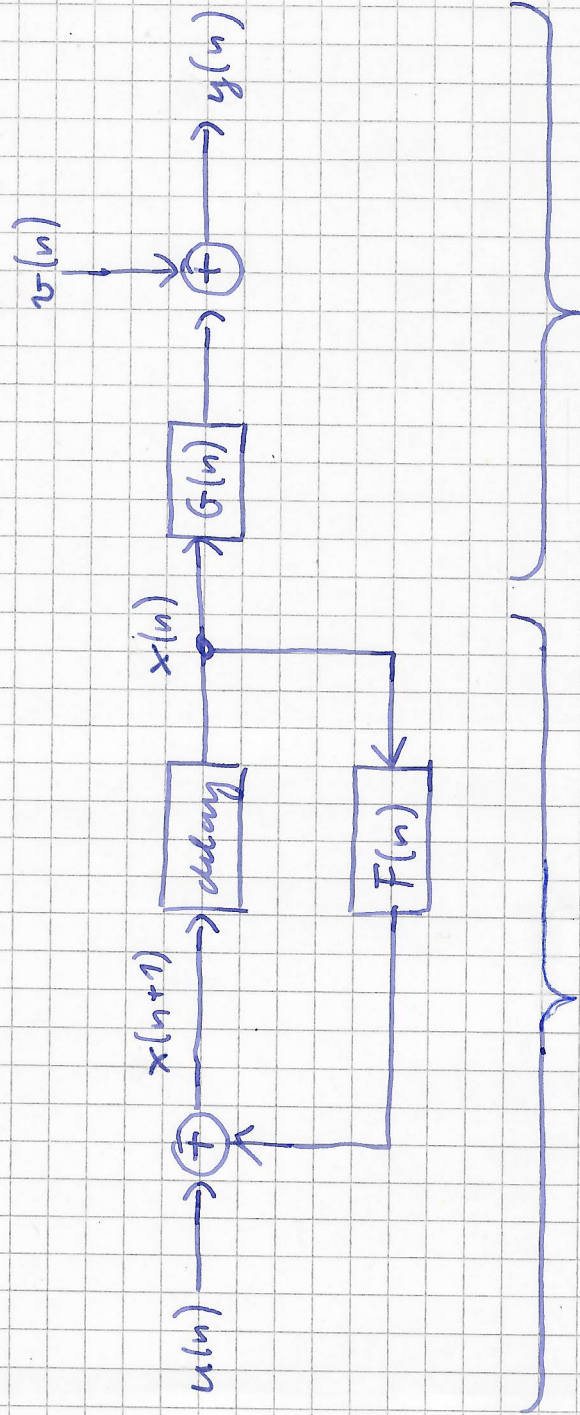
$$y(n) = G(n) \cdot x(n) + v(n)$$



zero mean observation noise
correlation matrix

$$R_{vv}(n) = E\{v(n) \cdot v^T(n)\}$$

Block Diagram of the Linear Dynamical System Model



State transition equation observation equation

Kinematic Model for Target Tracking

- for simplicity 1D-scenario
- sampling interval T
- second order model:
piecewise constant white acceleration model
- equations of motion

$$v_x(t+T) = v_x(t) + a_x(t) T$$

$$x(t+T) = x(t) + v_x(t) T + \frac{1}{2} a_x(t) T^2$$

- state transition equation

$$\begin{pmatrix} x(t+T) \\ v_x(t+T) \end{pmatrix} = \begin{pmatrix} 1 & T \\ 0 & 1 \end{pmatrix} \cdot \begin{pmatrix} x(t) \\ v_x(t) \end{pmatrix} + \begin{pmatrix} \frac{1}{2} a_x(t) T^2 \\ a_x(t) T \end{pmatrix}$$

$$x(n+1) = F(n) \cdot x(n) + u(n)$$

- correlation matrix of state noise

$$R_{nn}(n) = E\{u(n) \cdot u^T(n)\}$$

$$= E\{a_x^2(t)\} \begin{pmatrix} \frac{1}{4} T^4 & \frac{1}{2} T^3 \\ \frac{1}{2} T^3 & T^2 \end{pmatrix}$$