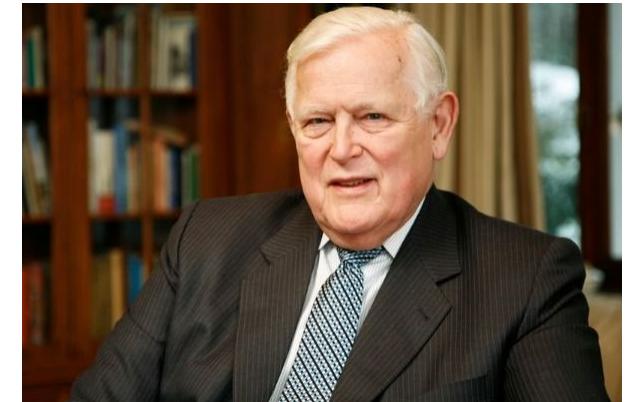


Advanced and Multivariable Control

Kalman Filter

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System

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t) + v_x(t) \\ y(t) &= Cx(t) + v_y(t)\end{aligned}$$

Disturbances

$$v = \begin{bmatrix} v_x \\ v_y \end{bmatrix} \xrightarrow{\text{white gaussian noise}} E[v(t)] = 0, \quad E[v(t_1)v(t_2)'] = V\delta(t_1 - t_2) \quad \delta \text{ Kroenecker index}$$

$$V = \begin{bmatrix} \tilde{Q} & Z \\ Z' & \tilde{R} \end{bmatrix}, \quad \tilde{Q} \in R^{n,n}, \quad \tilde{R} \in R^{p,p}, \quad Z \in R^{n,p}$$

$$\tilde{Q} \geq 0$$

state noise

$$\tilde{R} > 0$$

measurement noise

usually
 $\neq 0$

for instance some outputs
and states can coincide

Initial assumption (to be removed) $\longrightarrow Z = 0$

Initial state

$$x_0 = x(0) \xrightarrow{\text{Gaussian random variable}} E[x_0] = \bar{x}_0 , \quad E[(x_0 - \bar{x}_0)(x_0 - \bar{x}_0)'] = \tilde{P}_0 \geq 0$$

Additional assumption

$$E[x_0 [v'_x \quad v'_y]] = 0$$

Goal: to design a state observer «optimal» according to a given criterion which considers the *presence and characteristics* of the noises

otherwise a pole placement observer could be used



Some comments on the parameters

Proper knowledge of \bar{x}_0, \tilde{P}_0 , not so critical

Knowledge of $\tilde{R} > 0$: possible, it represents the measurement noise, often provided by sensors' manufactures

Knowledge of $\tilde{Q} \geq 0$: often difficult. Can we use it as a design parameter to obtain faster or slower state estimation? See later



Structure of the filter/observer

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t) + v_x(t) \\ y(t) &= Cx(t) + v_y(t)\end{aligned}$$

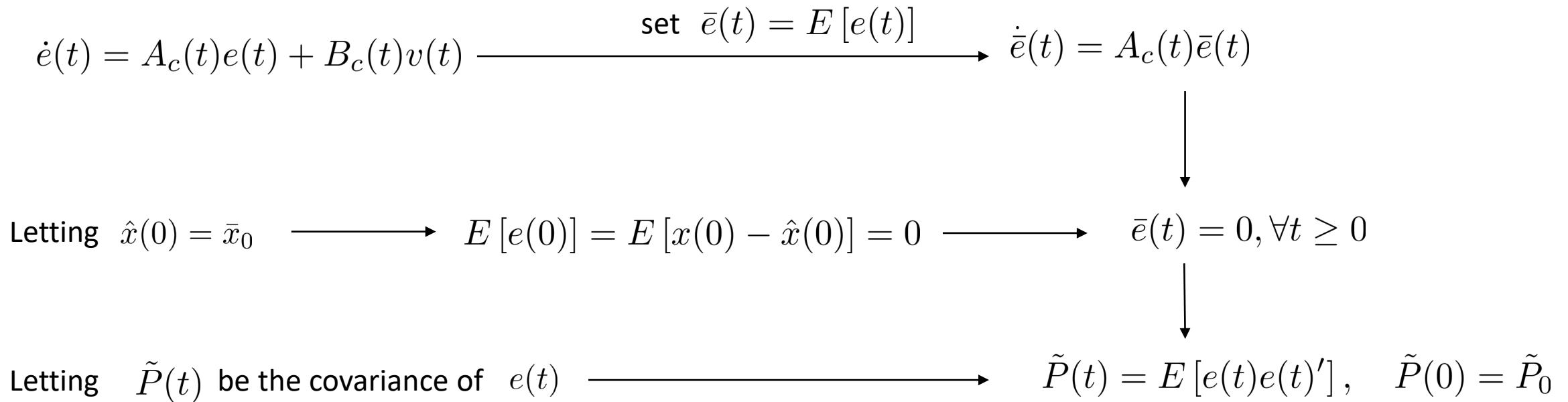
$$\dot{\hat{x}}(t) = A\hat{x}(t) + Bu(t) + L(t)[y(t) - C\hat{x}(t)]$$

time varying gain to be selected
according to an optimality criterion

State estimation error $e(t) = x(t) - \hat{x}(t)$

$$\begin{aligned}\dot{e}(t) &= Ae(t) - L(t)[Ce(t) + v_y(t)] + v_x(t) \\ &= [A - L(t)C]e(t) + B_c(t)v(t) \\ &= A_c(t)e(t) + B_c(t)v(t)\end{aligned}$$

$$A_c(t) = [A - L(t)C], \quad B_c(t) = [I \quad -L(t)]$$



Goal: minimize with respect to the gain $L(t)$ the covariance of the estimation error

$$\min_{L(t)} \gamma' \tilde{P}(t) \gamma$$

$\gamma \in R^{n,1}$ is a generic vector

The solution of the above problem is

$$L(t) = \tilde{P}(t)C'R^{-1}$$

where $\tilde{P}(t)$ is the solution of the Riccati equation

$$\dot{\tilde{P}}(t) = A\tilde{P}(t) + \tilde{P}(t)A' + \tilde{Q} - \tilde{P}(t)C'R^{-1}C\tilde{P}(t)$$

with initial condition

$$\tilde{P}(0) = \tilde{P}_0$$

Matrix differential Riccati equation to be solved

LQ - KF

$$\min_{K(t)} x' P(t) x$$

$$K(t) = R^{-1} B' P(t)$$

$$\dot{P}(t) + Q - P(t)BR^{-1}B'P'(t) + P(t)A + A'P'(t) = 0$$

$$P(T) = S$$

$$\xleftarrow{t}$$

$$\min_{L(t)} \gamma' \tilde{P}(t) \gamma$$

$$L(t) = \tilde{P}(t) C' \tilde{R}^{-1}$$

$$\dot{\tilde{P}}(t) = A\tilde{P}(t) + \tilde{P}(t)A' + \tilde{Q} - \tilde{P}(t)C'\tilde{R}^{-1}C\tilde{P}(t)$$

$$\tilde{P}(0) = \tilde{P}_0$$

$$\xrightarrow{t}$$

LQ KF

A A'

B C'

Q \tilde{Q}

R \tilde{R}

P \tilde{P}

t $t^ - t$*

K L'

*perfect duality previous results on LQ can
be extended to KF*

KF with time varying gain $L(t)$: not very useful and difficult to implement. Is it possible to compute a steady-state filter?

In view of duality, we can extend the results of LQ_{inf}

If

1. the pair (A, B_q) is reachable, where B_q is such that $\tilde{Q} = B_q B'_q$;
2. the pair (A, C) is observable;

then

- A) the optimal estimator is

$$\begin{aligned}\dot{\hat{x}}(t) &= A\hat{x}(t) + Bu(t) + \bar{L}[y(t) - C\hat{x}(t)] \\ &= (A - \bar{L}C)\hat{x}(t) + Bu(t) + \bar{L}y(t)\end{aligned}$$

with

$$\bar{L} = \bar{P}C'\tilde{R}^{-1}$$

where \bar{P} is the unique positive definite solution of the stationary Riccati equation

$$0 = A\bar{P} + \bar{P}A' + \tilde{Q} - \bar{P}C'\tilde{R}^{-1}C\bar{P}$$

- B) the observer is asymptotically stable, that is all the eigenvalues of $(A - \bar{L}C)$ have negative real part.

What to do if $Z \neq 0$?

Write the system as

$$\begin{aligned}
 \dot{x}(t) &= Ax(t) + Bu(t) + v_x(t) + Z\tilde{R}^{-1}(y(t) - y(t)) \\
 &= Ax(t) + Bu(t) + v_x(t) + Z\tilde{R}^{-1}y(t) - Z\tilde{R}^{-1}Cx(t) \\
 &\quad - Z\tilde{R}^{-1}v_y(t) \\
 &= \underbrace{\left(A - Z\tilde{R}^{-1}C \right)}_{\bar{A}} x(t) + Bu(t) + Z\tilde{R}^{-1}y(t) + \bar{v}(t)
 \end{aligned}$$

$\bar{v}(t) = v_x(t) - Z\tilde{R}^{-1}v_y(t)$
 $E[\bar{v}(t)v_y(t)'] = Z - Z\tilde{R}^{-1}\tilde{R} = 0$
 $E[\bar{v}(t)\bar{v}(t)'] = \tilde{Q} - Z\tilde{R}^{-1}Z' = \tilde{Q} \geq 0$



$$\begin{aligned}
 \dot{x}(t) &= \bar{A}x(t) + Bu(t) + Z\tilde{R}^{-1}y(t) + \bar{v}(t) \\
 y(t) &= Cx(t) + v_y(t)
 \end{aligned}$$

Satisfies the initial assumptions of KF

use it as a standard input

Duality with LQ allows one to understand how to choose the design parameters (provided that they are not obtained from physical considerations)

 LQ_{inf}

Q "large", R "small" → "fast" feedback system

 $KF_{stationary}$

\tilde{Q} "large", \tilde{R} "small" → "fast" observer

and viceversa



Example: estimation of a constant from noisy measurements

$$\begin{aligned}\dot{x}(t) &= 0, \quad x(0) = \bar{x} \\ y(t) &= x(t) + v_y(t)\end{aligned}$$

$$A = 0, C = 1, \tilde{Q} = 0, \tilde{R} = r$$

$(A, \tilde{Q}^{1/2})$ **not reachable**

Riccati equation

$$\frac{d\tilde{P}(t)}{dt} = -\frac{\tilde{P}^2(t)}{r} \longrightarrow \tilde{P}(t) = \frac{1}{r^{-1}t + \frac{1}{\tilde{P}_0}} \longrightarrow \tilde{P}(t) \rightarrow 0$$

$$L(t) = \tilde{P}(t)C\tilde{R}^{-1} = \frac{r^{-1}}{r^{-1}t + \frac{1}{\tilde{P}_0}}$$

The KF gain tends to 0, the estimate tends to be constant

$$\dot{\hat{x}}(t) = \frac{r^{-1}}{r^{-1}t + \frac{1}{\tilde{P}_0}}(y(t) - \hat{x}(t))$$

Example

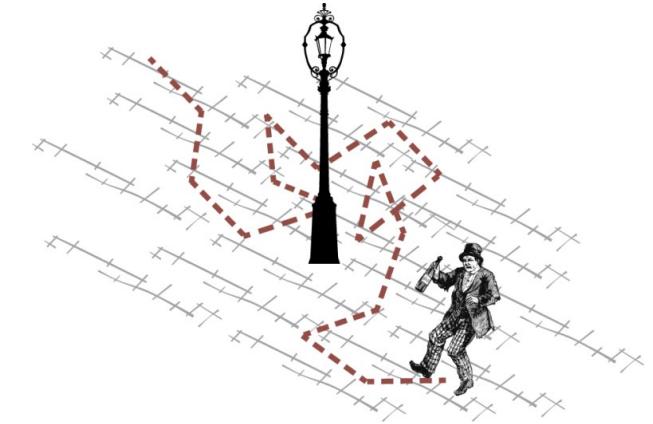
$$\begin{aligned}\dot{x}(t) &= Ax(t) + v_x(t) \\ y(t) &= v_y(t)\end{aligned} \longrightarrow C = 0 \longrightarrow L(t) = 0 \longrightarrow \dot{\hat{x}}(t) = A\hat{x}(t)$$

*The output does not bring any information on the state
You can only run the estimator open-loop from the knowledge of the expected value of the initial state*

Example

$$\begin{aligned}\dot{x}(t) &= v_x(t) \\ y(t) &= Cx(t) + v_y(t)\end{aligned}$$

*Brownian motion, or
drunkard's Walk*



Steady state Riccati equation $0 = \tilde{Q} - \frac{\bar{P}^2 C^2}{\tilde{R}} \longrightarrow \bar{P} = \sqrt{\frac{\tilde{Q}\tilde{R}}{C^2}} \longrightarrow \bar{L} = \bar{P}C'\tilde{R}^{-1} = \sqrt{\frac{\tilde{Q}\tilde{R}}{C^2}} \frac{C}{\tilde{R}} = \sqrt{\frac{\tilde{Q}}{\tilde{R}}}$

$$\frac{d\hat{x}(t)}{dt} = \sqrt{\frac{\tilde{Q}}{\tilde{R}}} [y(t) - C\hat{x}(t)] \quad \left\{ \begin{array}{l} \frac{\tilde{Q}}{\tilde{R}} \text{ large: "fast observer"} \\ \frac{\tilde{Q}}{\tilde{R}} \text{ small: "slow observer"} \end{array} \right.$$

$$\hat{X}(s) = \frac{1}{1 + \frac{1}{C} \sqrt{\frac{\tilde{R}}{\tilde{Q}}} s} Y(s)$$

Estimation of a Brownian motion

$$\begin{cases} \dot{x} = v_x \\ y_t = x + v_y \end{cases}$$

$v_x \sim \text{WGN}(0, 1)$, $v_y \sim \text{WGN}(0, 1)$

$$A=0, C=1$$

Riccati eq. $\rightarrow 1 - \tilde{P}^2 = 0 \rightarrow \tilde{P} = 1$ asymptotic variance of
the state estimation
error

$$\tilde{P}=1 \rightarrow L = \tilde{P} C' \tilde{R}^{-1} = 1$$

Observe $\dot{\hat{x}} = [y_t - \hat{x}] \rightarrow \frac{y_t}{s+1} \rightarrow \hat{x}$

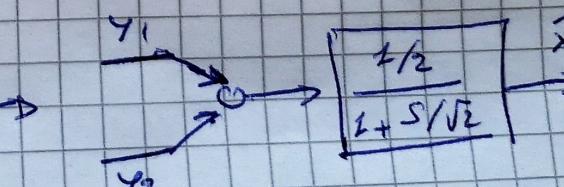
Now assume to have two measurements with the same measurement noise

$$\begin{cases} \overset{\circ}{x} = v_x \\ y_1 = x + v_{y_1} \\ y_2 = x + v_{y_2} \end{cases} \quad \tilde{Q} = I, \quad \tilde{R} = I_2, \quad A = 0, \quad C = \begin{vmatrix} 1 & 1 \\ 1 & 1 \end{vmatrix}$$

Riccati eq. $I - \tilde{P} \begin{vmatrix} 1 & 1 \\ 1 & 1 \end{vmatrix} I = \begin{vmatrix} 1 & 1 \\ 1 & 1 \end{vmatrix} \tilde{P} = 0 \rightarrow I - 2\tilde{P}^2 = 0$

$$\tilde{P} = \sqrt{\frac{I}{2}} \approx 0.707 \quad \text{were variance of the state estimation error}$$

$$L = \frac{I}{\sqrt{2}} \begin{vmatrix} 1 & 1 \\ 1 & 1 \end{vmatrix}$$

$$\begin{aligned} \overset{\circ}{x} &= \begin{vmatrix} 1 & 1 \\ 1 & 1 \end{vmatrix} \frac{I}{\sqrt{2}} \left\{ \begin{vmatrix} y_1 \\ y_2 \end{vmatrix} - \begin{vmatrix} 1 & 1 \\ 1 & 1 \end{vmatrix} \hat{x} \right\} \\ &= \frac{I}{\sqrt{2}} \left\{ y_1 + y_2 - 2\hat{x} \right\} \end{aligned}$$


Now assume to have 2 measurements with different measurement noise

$$\tilde{Q} = I \quad \tilde{R} = \begin{vmatrix} 1 & 0 \\ 0 & 2 \end{vmatrix}$$

Riccati eq. $I - \tilde{P} \begin{vmatrix} 1 & 1 \\ 0 & \frac{1}{2} \end{vmatrix} \begin{vmatrix} 1 & 0 \\ 0 & \frac{1}{2} \end{vmatrix} \begin{vmatrix} 1 \\ 2 \end{vmatrix} \tilde{P} = 0$

it is easy to compute $\tilde{P} = \sqrt{\frac{2}{2+1}} \rightarrow L = \sqrt{\frac{2}{2+1}} \begin{vmatrix} 1 & \frac{1}{2} \end{vmatrix}$

$$\hat{x}^* = \sqrt{\frac{2}{2+1}} \begin{vmatrix} 1 & \frac{1}{2} \end{vmatrix} \begin{vmatrix} y_1 - \hat{x}^* \\ y_2 - \hat{x}^* \end{vmatrix}$$

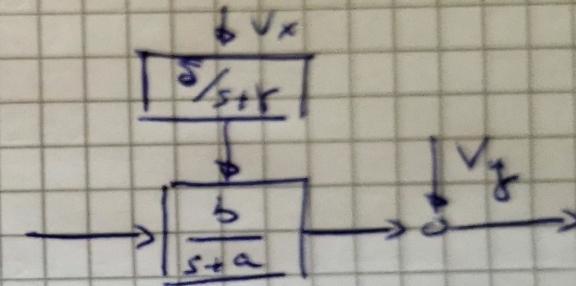
$$\hat{x} = \sqrt{\frac{r}{r+1}} \left\{ (y_1 - \hat{x}) + \frac{1}{r} (y_2 - \hat{x}) \right\}$$

If $r \rightarrow \infty$ $\hat{x} = y_1 - \hat{x}$ The second measurement y_2
is not reliable and
not used

If $r \rightarrow 0$ only the second measurement is
used because it is not affected
by measurement noise

KF with non white noise

$$\begin{cases} \dot{x} = -ax + bu + v \\ y = x + v_y \end{cases}, \quad \dot{v} = -\gamma v + \sigma v_x, \quad \gamma > 0$$



$$\begin{cases} \begin{pmatrix} \dot{x} \\ \dot{v} \end{pmatrix} = \underbrace{\begin{pmatrix} -a & 1 \\ 0 & -\gamma \end{pmatrix}}_A \begin{pmatrix} x \\ v \end{pmatrix} + \underbrace{\begin{pmatrix} b \\ 0 \end{pmatrix}}_B u + \underbrace{\begin{pmatrix} 0 \\ \sigma \end{pmatrix}}_{\sim} v_x \\ y = \underbrace{\begin{pmatrix} 1 & 0 \end{pmatrix}}_C \begin{pmatrix} x \\ v \end{pmatrix} + v_y \end{cases}$$

$$\mathbb{E}[\tilde{v}_x] = 0, \quad \mathbb{E}[\tilde{v}_x'] = 0$$

$$\mathbb{E}[\tilde{v}_x \tilde{v}_x'] = H \mathbb{E}[v_x v_x'] H' = \underbrace{H \tilde{\varphi}_x H'}_{H q^2 H'} \quad (\tilde{\varphi}_x = q^2 I)$$

$$= \begin{bmatrix} 0 & 0 \\ 0 & q^2 \delta^2 \end{bmatrix} \underbrace{\tilde{\varphi}}_{\tilde{\varphi} \geq 0}$$

$$B_q B_q' = \begin{vmatrix} 0 & 1 \\ \tilde{s}q & \end{vmatrix} \begin{vmatrix} 0 & \tilde{s}q \\ 1 & \end{vmatrix}$$

$$\text{Observability } (A, c) \rightarrow M_0 = \left| \begin{matrix} c \\ cA \end{matrix} \right| = \begin{vmatrix} 1 & 0 \\ -a & 1 \end{vmatrix} \neq 0$$

$$\text{Reachability } (A, B_q) \rightarrow M_2 = \left| \begin{matrix} B_q & AB_q \end{matrix} \right| = \begin{vmatrix} 0 & \tilde{s}q \\ \tilde{s}q & -q \tilde{s}q \end{vmatrix} \xrightarrow{\text{rank}} \begin{array}{l} \tilde{s} \neq 0 \\ q \neq 0 \end{array}$$

Design of the filter for the enlarged state

$$\tilde{\tilde{X}} = \begin{vmatrix} X \\ V \end{vmatrix}$$

$$\dot{\tilde{\tilde{X}}} = A\tilde{\tilde{X}} + Bu + L [y - c\tilde{\tilde{X}}] , L = \begin{vmatrix} l_1 \\ l_2 \end{vmatrix}$$

$$\left\{ \begin{array}{l} \dot{\tilde{X}} = -a\tilde{X} + \hat{V} + bu + l_1 (y - \hat{X}) \\ \dot{\hat{V}} = -\gamma \hat{V} + l_2 [y - \hat{X}] \end{array} \right.$$



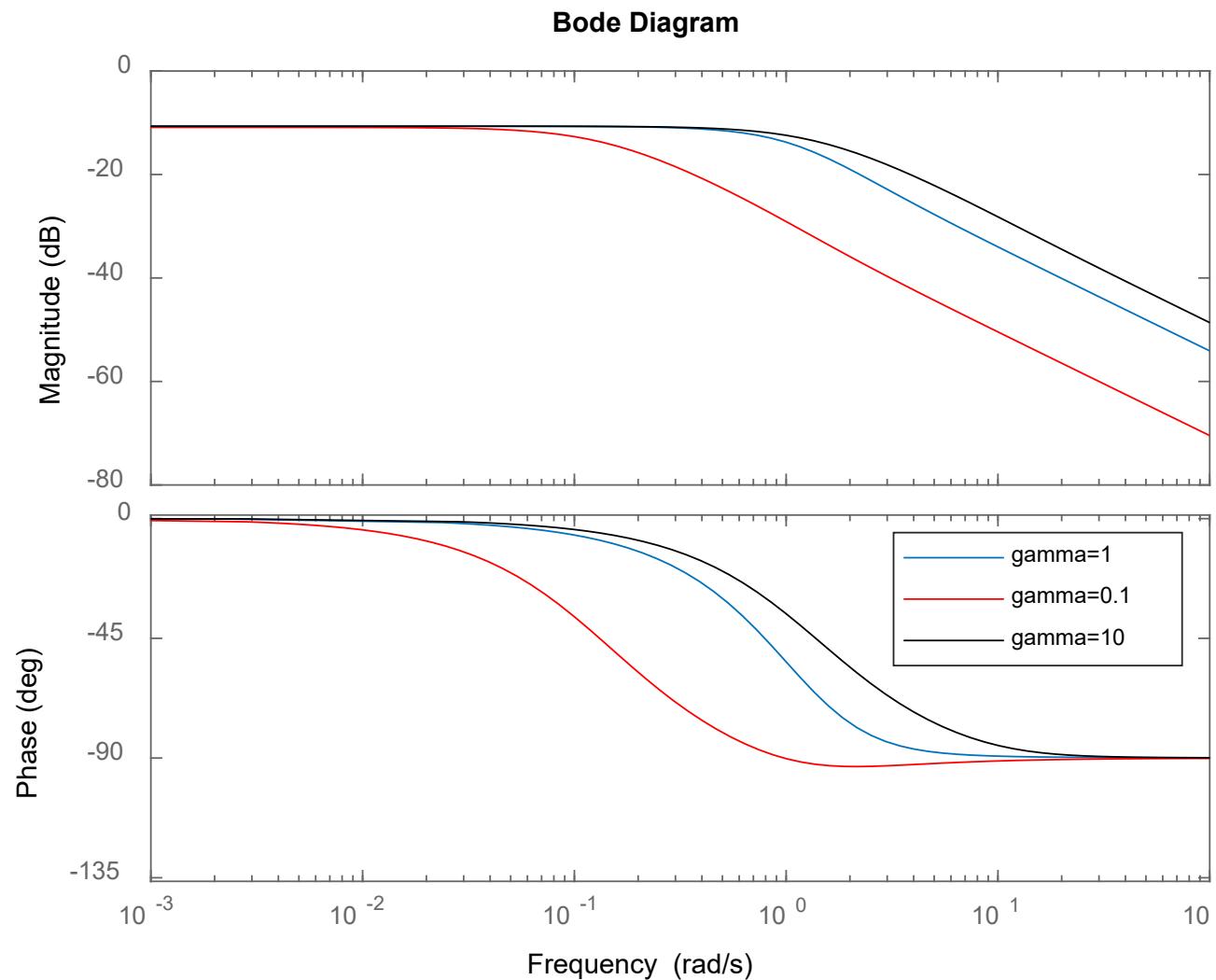
$$\hat{x}(s) = \underbrace{C}_{\text{Transfer function of}} \underbrace{(sI - (A - LC))^{-1} L}_{\text{the filter}} y(s)$$

Transfer function of
the filter

Considering fixed values

$$a=1, \tilde{R}=1, \tilde{\delta}=\gamma, \tilde{q}=1$$

we obtain different transfer functions of the filter



The slower the system of the noise, the slower the filter