

Control over Unreliable Communication Links

(based on joint work with O.C. Imer, S.Yüksel)

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Outline

- Networks & Control
- Issues in Networked Control Systems
 - What/when/how to transmit & control
- Control with Unreliable Channels
- Control with Power-Limited Communication
 - When to measure & when to control
- Conclusions

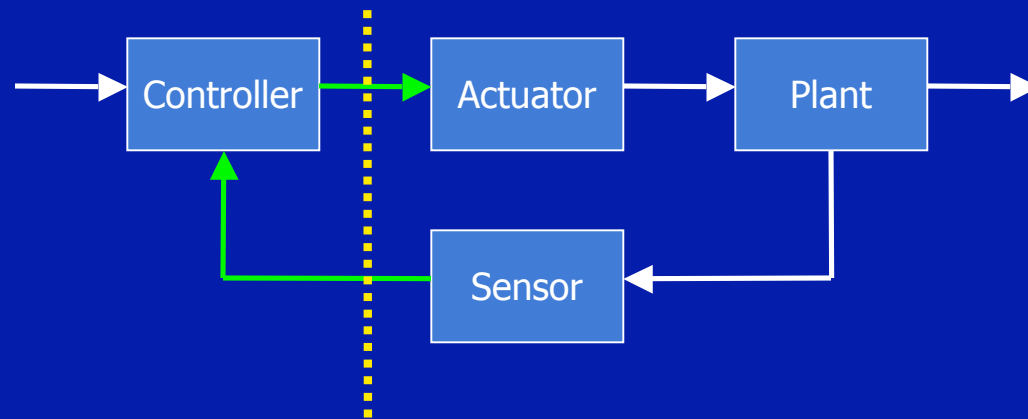
Networks & Control

- A surge of research interest in using networks for communication in control/monitoring systems
- **Advantages:** flexibility, reduced wiring, lower installation costs, agility in diagnosis and maintenance, remote operations, ...
- **Challenges:** synchronization, timing problems, reliability, *communication network and channel constraints*, ...

Networks & Control

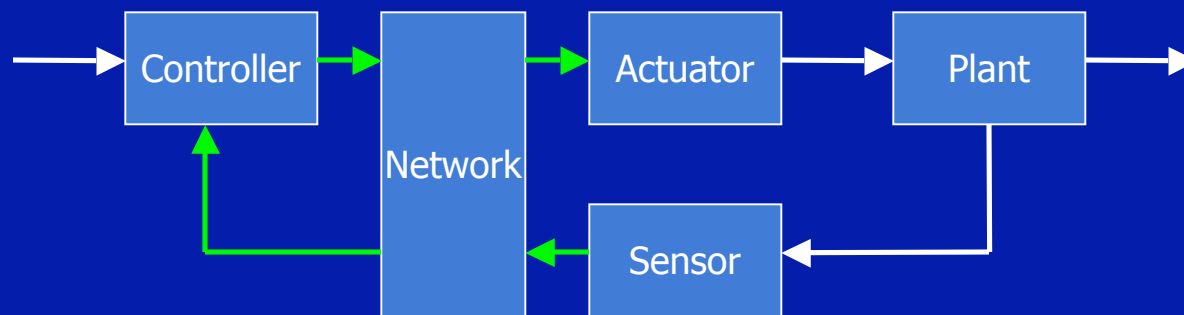
- Traditional Control Systems have **dedicated data paths** for communication

Controller-Plant Communication

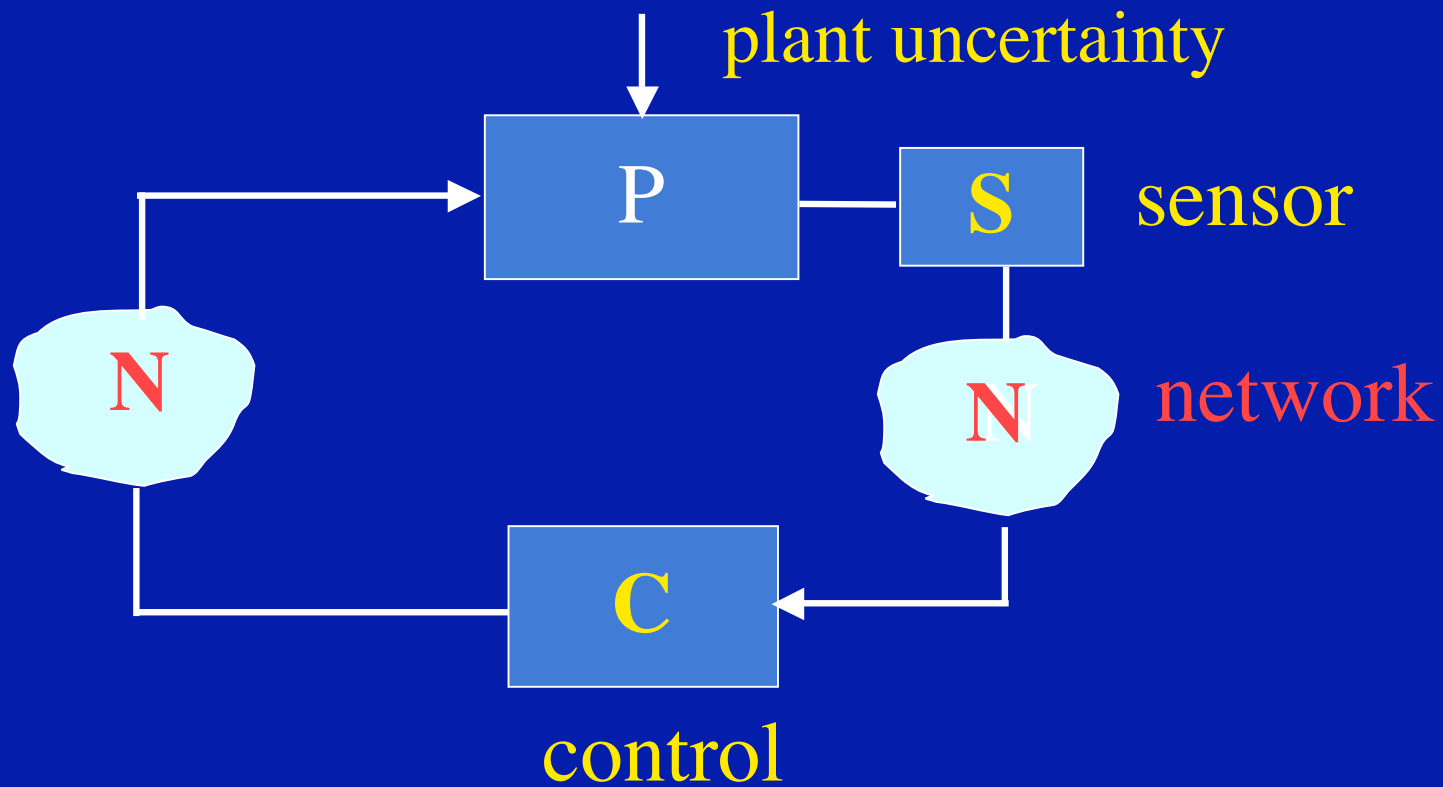


Networks & Control

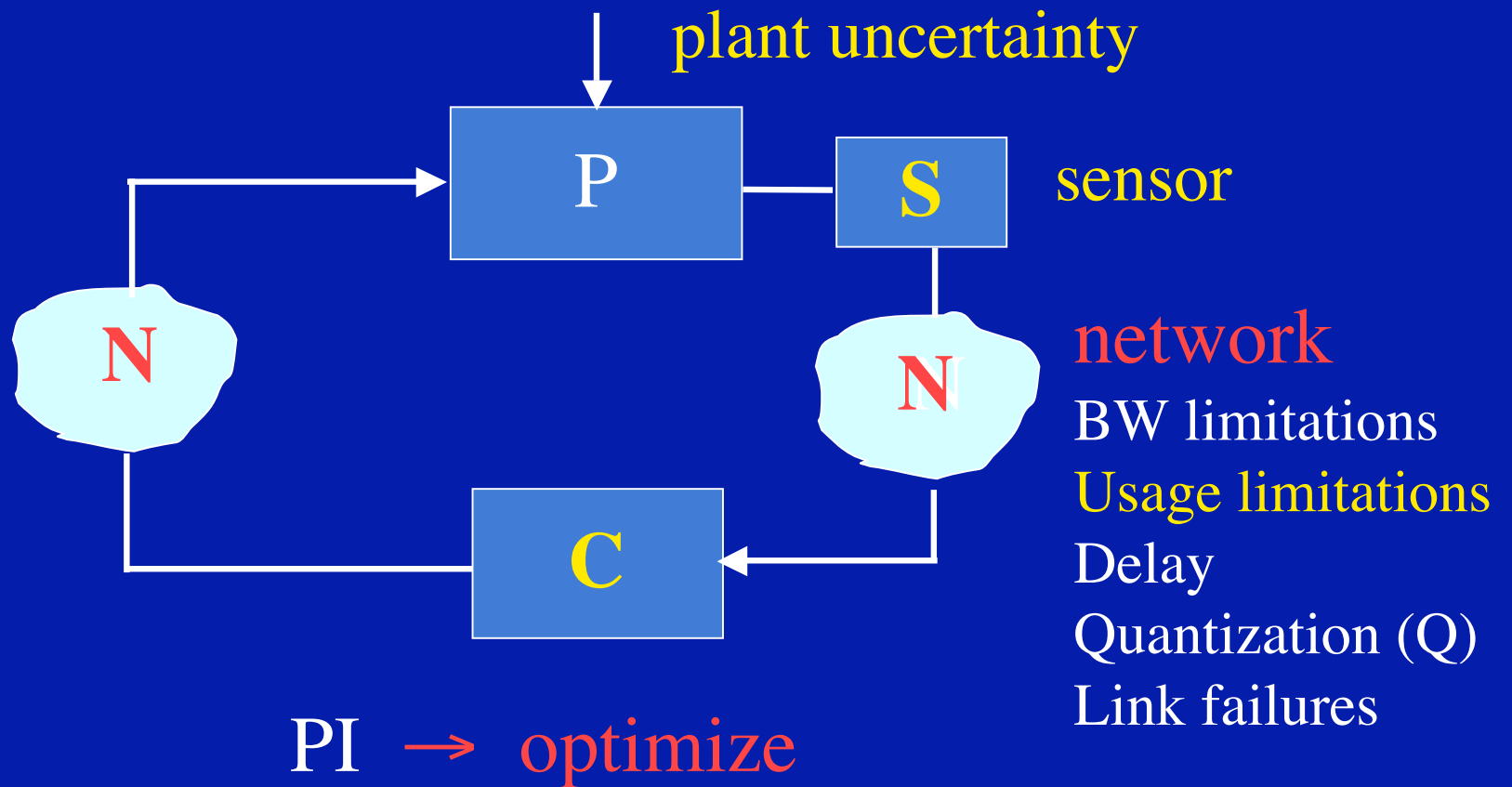
- In Networked Control Systems (NCS) **controller-plant communication** takes place over a **network**



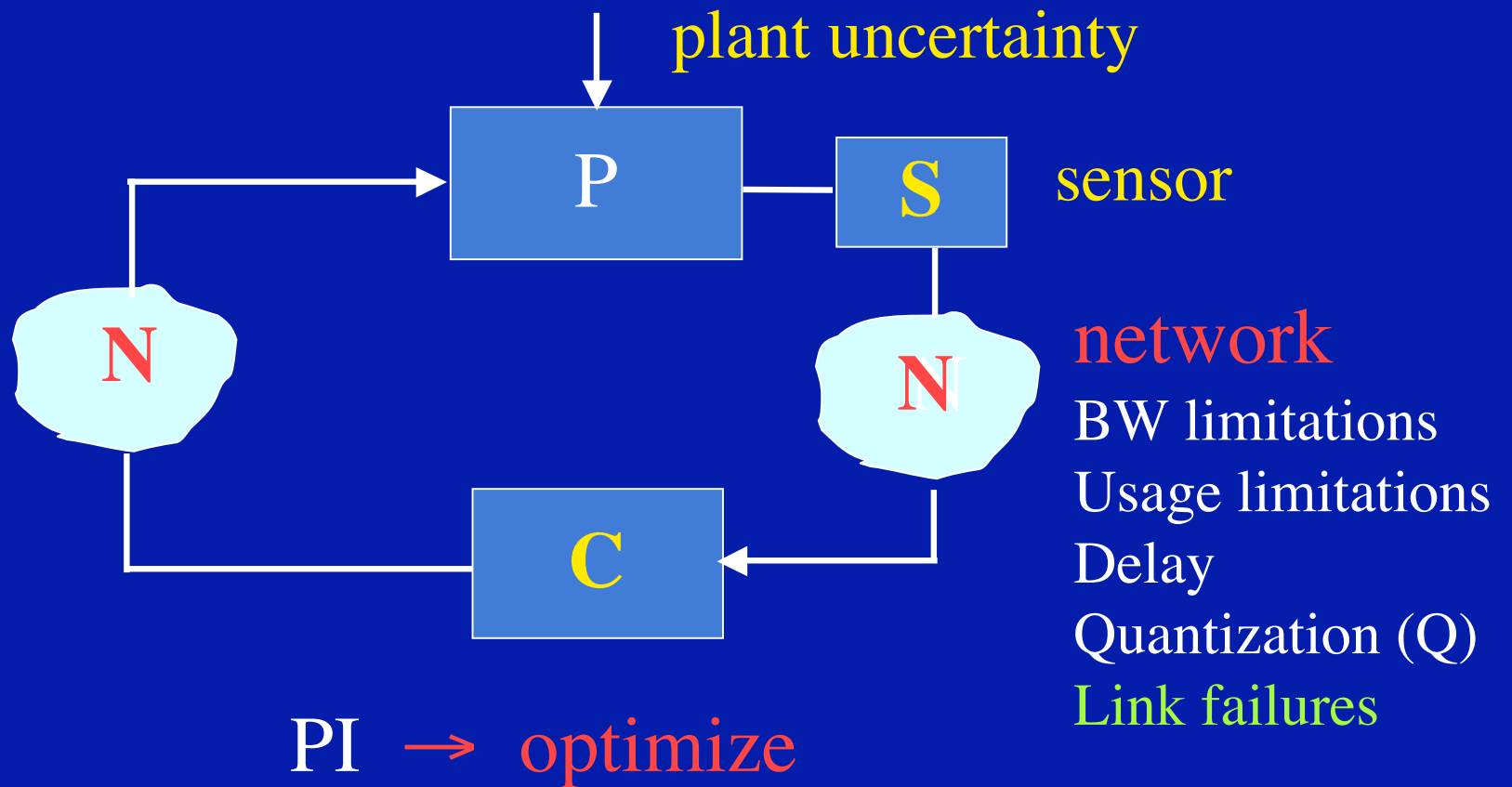
Remote Control Paradigm



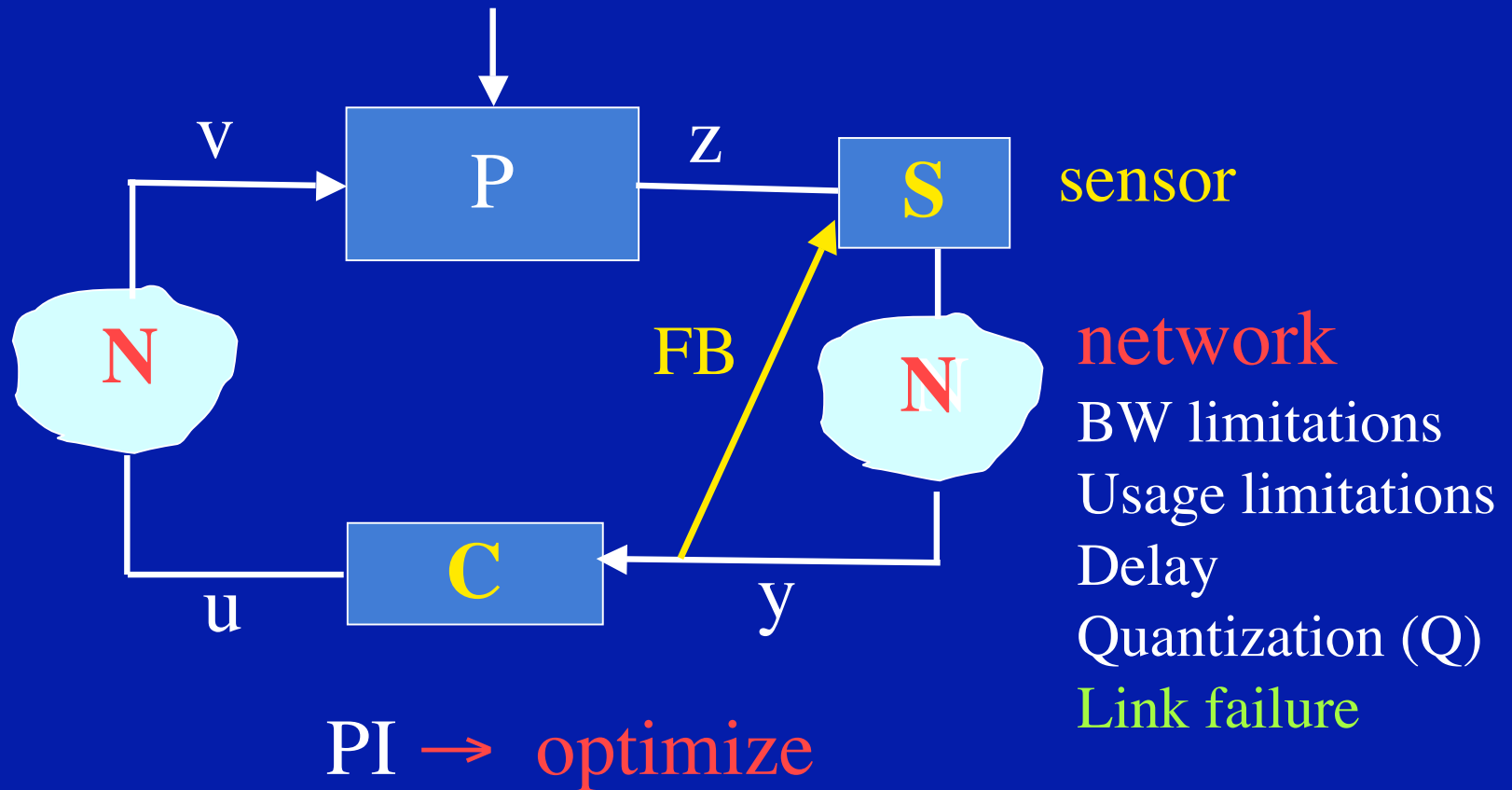
Remote Control Paradigm



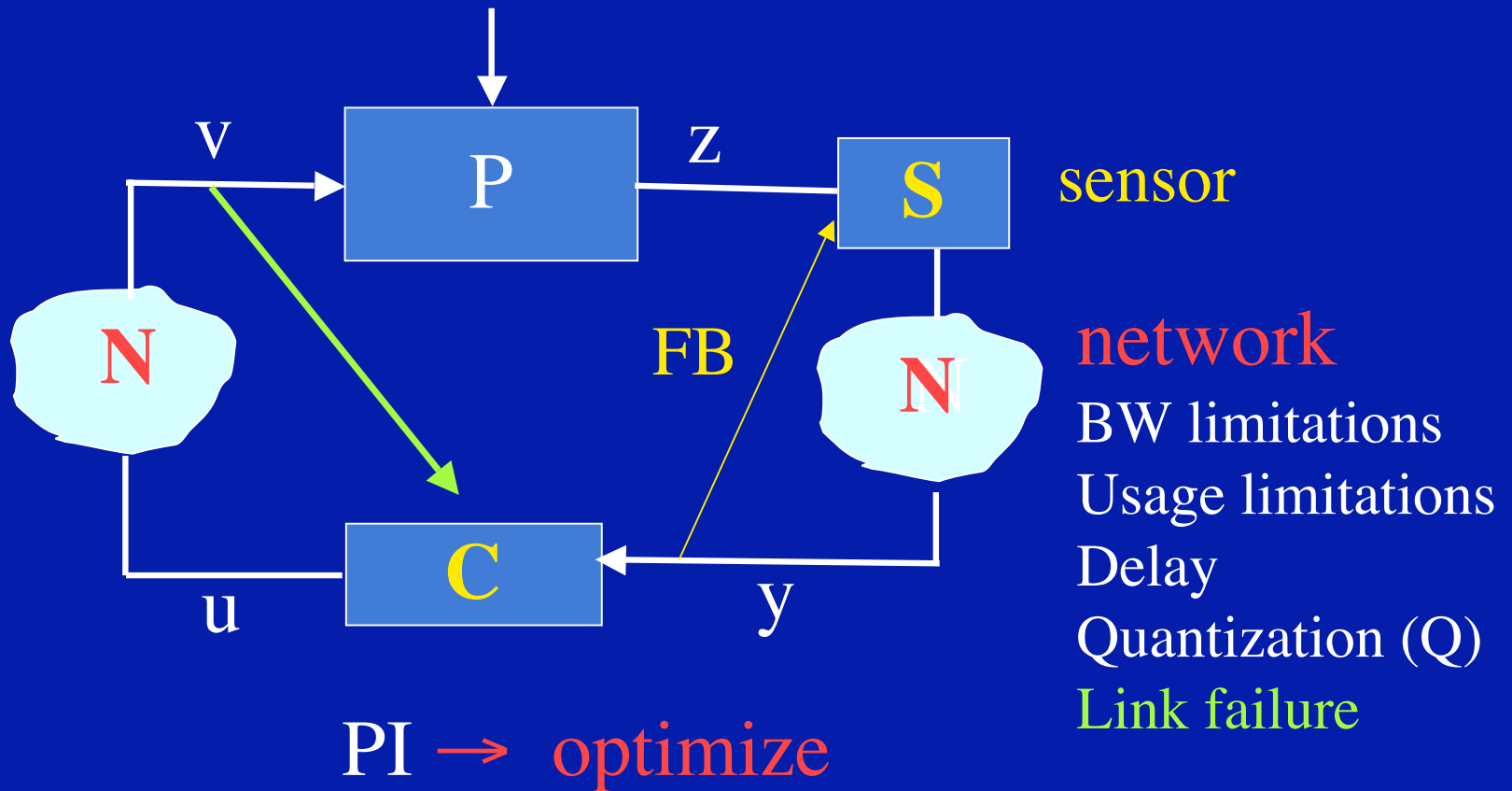
Remote Control Paradigm



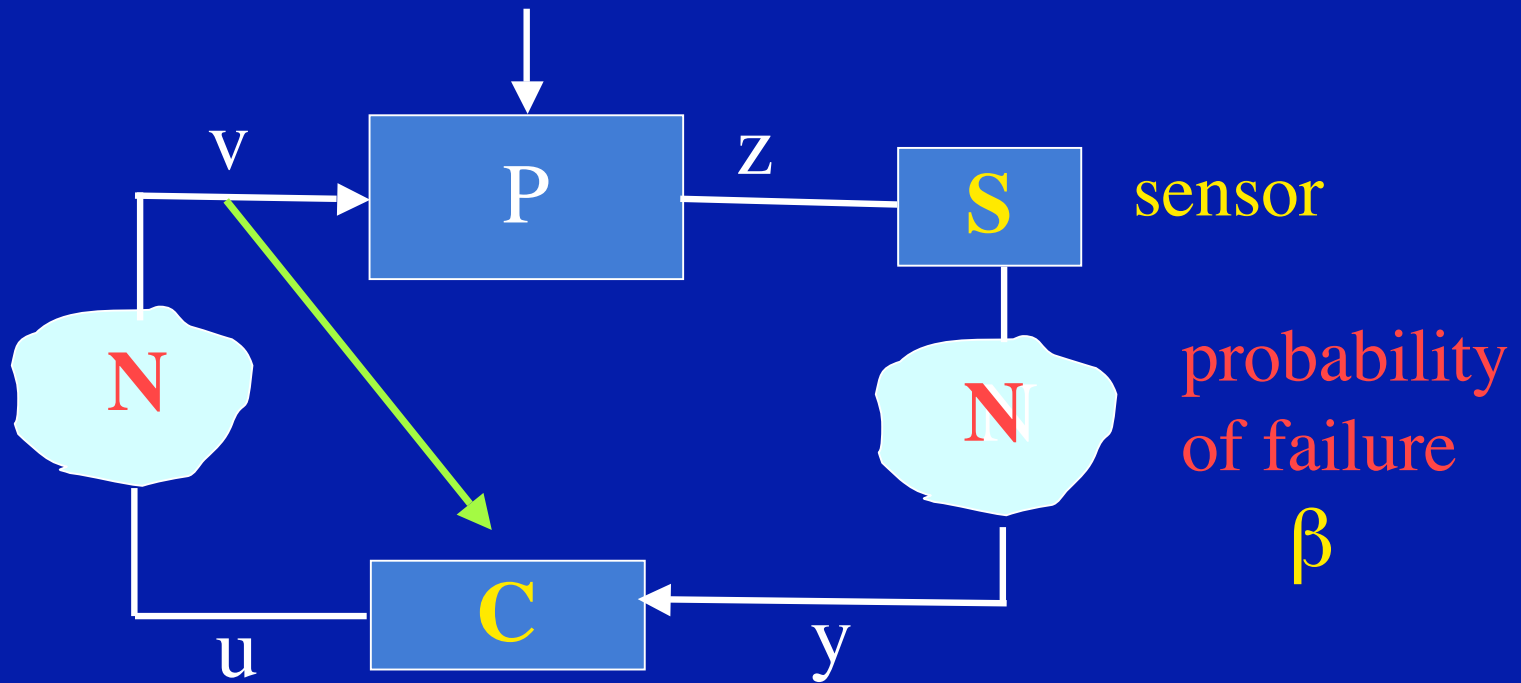
Remote Control Paradigm



Remote Control Paradigm

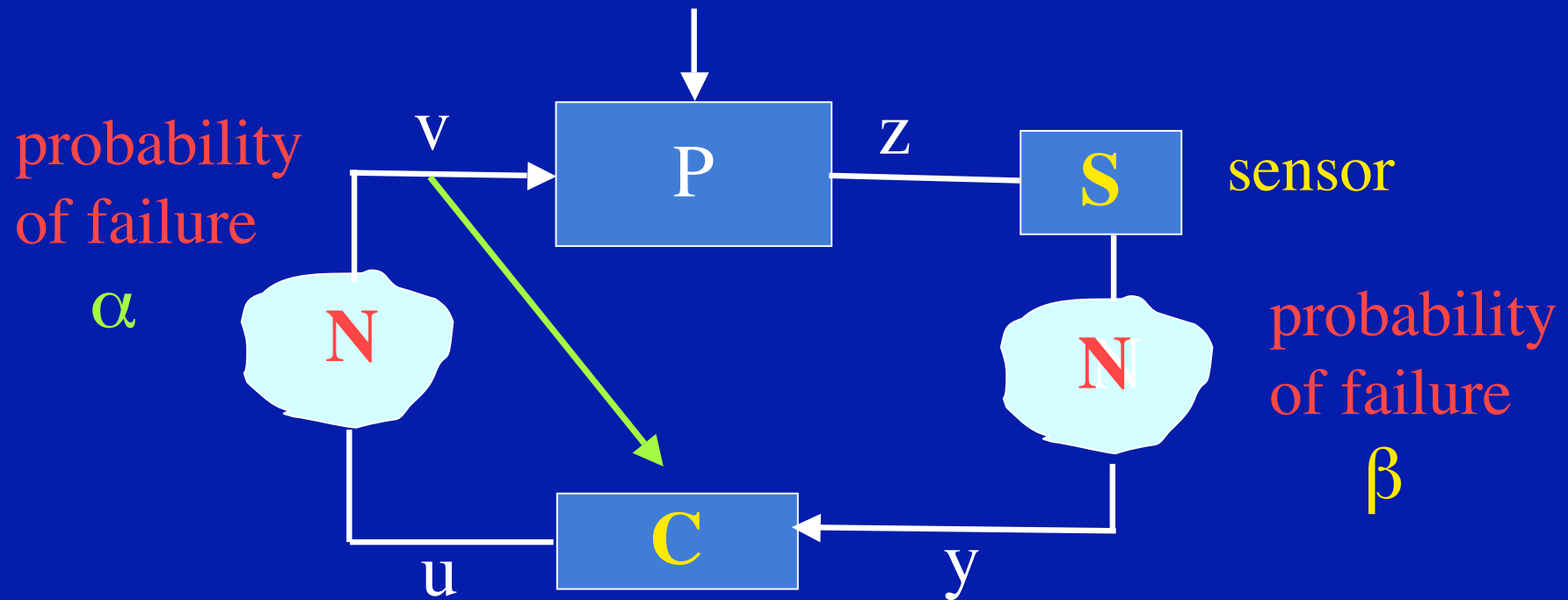


Link Failures



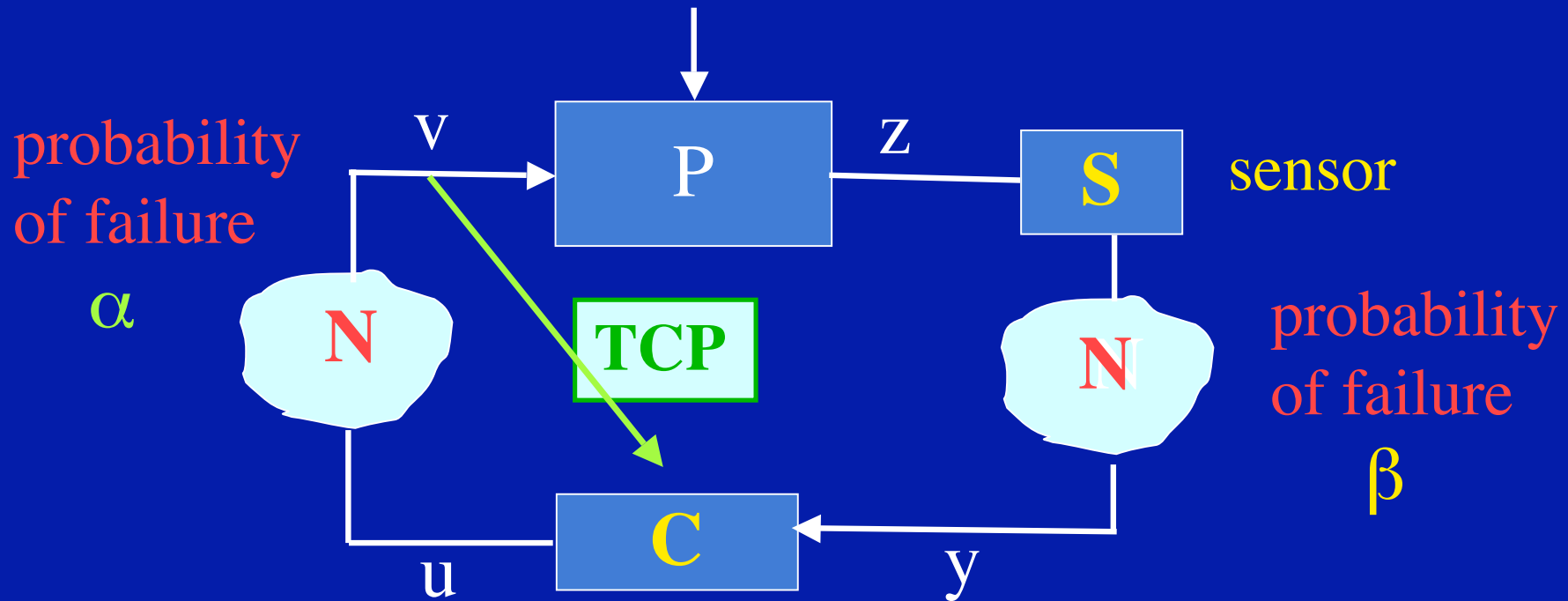
PI \rightarrow optimize

Link Failures



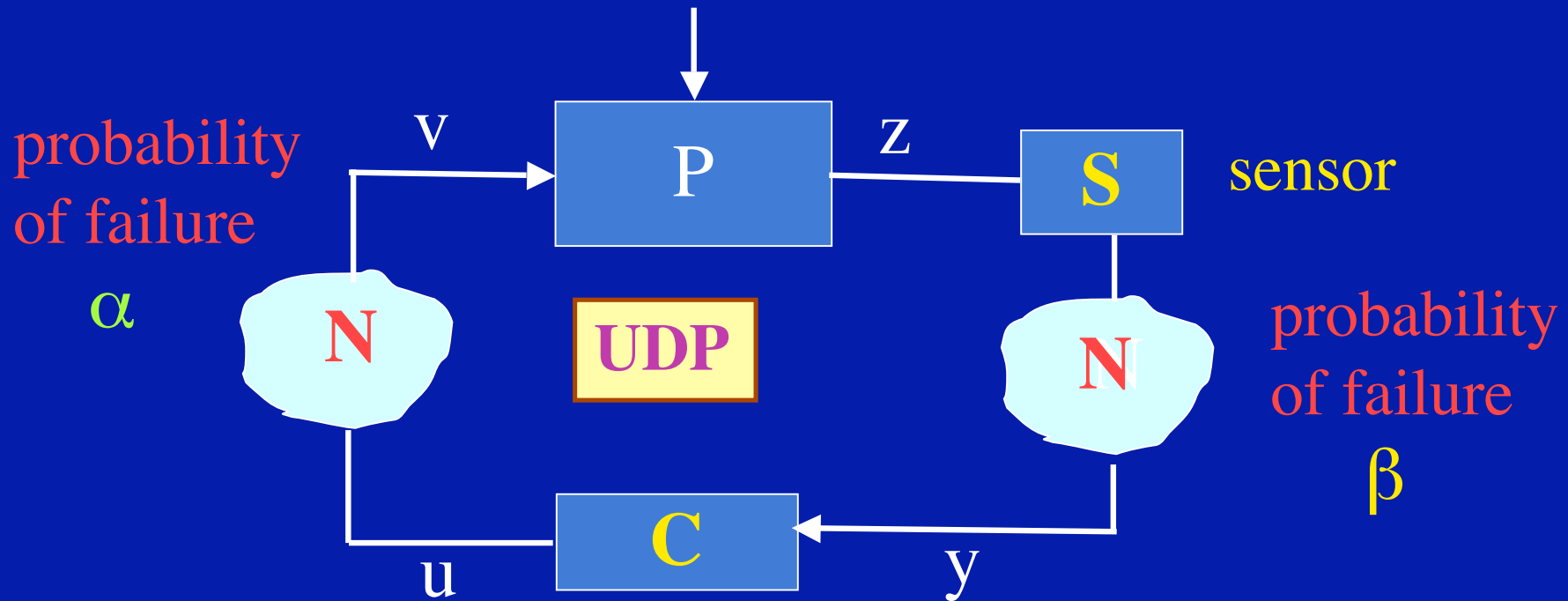
PI \rightarrow optimize

Scenario I



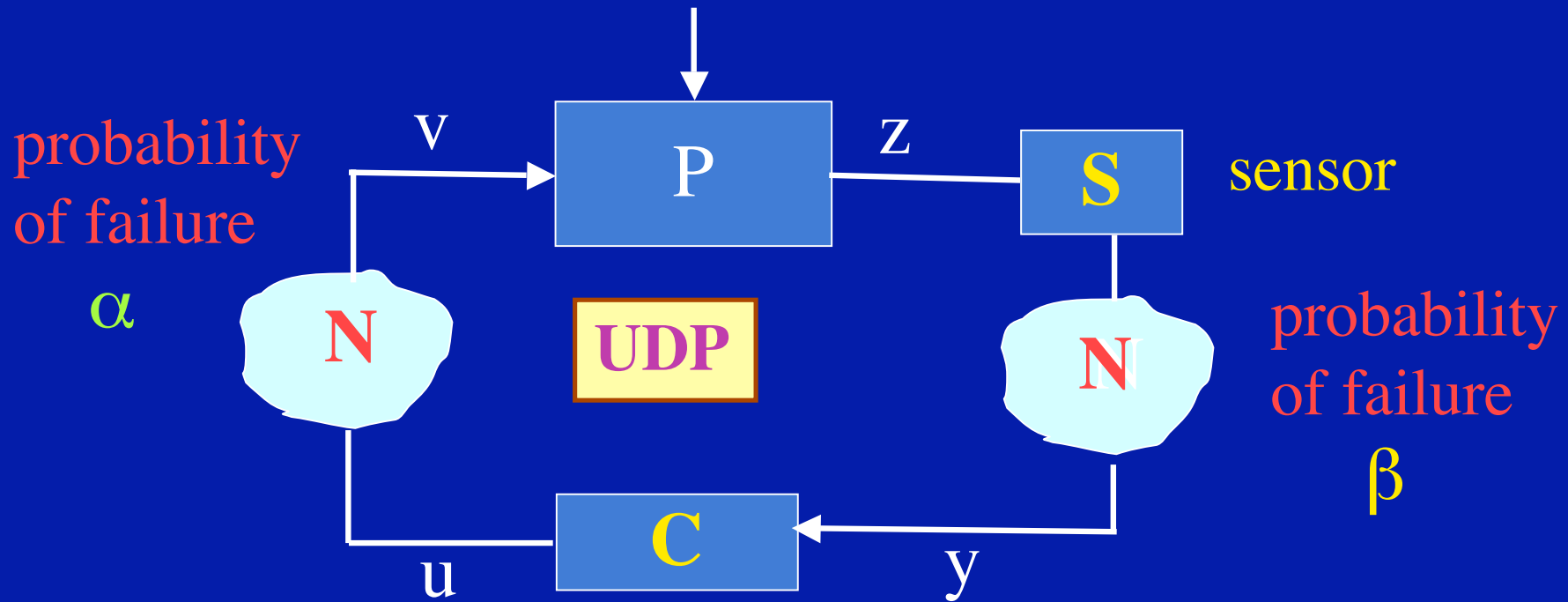
Transmission Control Protocol of Internet
(with acknowledgements)

Scenario II



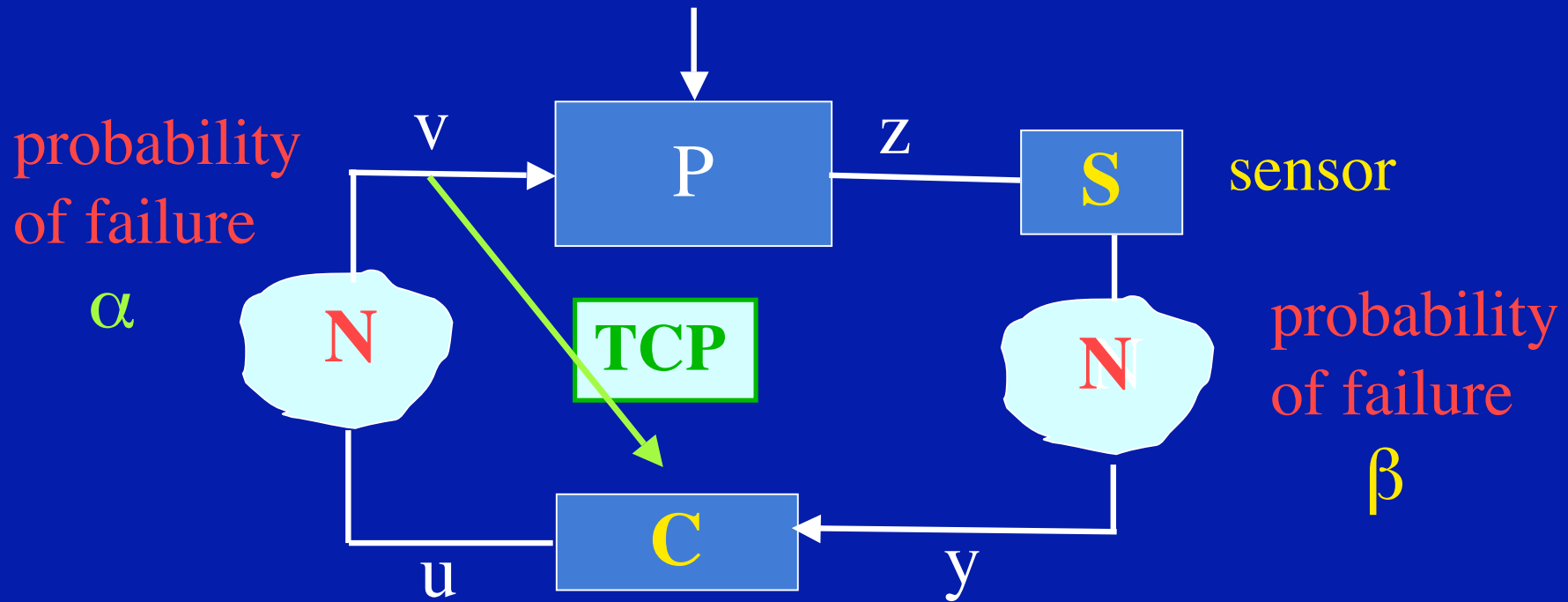
User Datagram Protocol -- best effort network
(no acknowledgements)

Scenario II



$$I_k^{\text{UDP}} = \{y_0^k, u_0^{k-1}, \beta_0^k\}$$
$$u_k = \mu_k(I_k^{\text{UDP}})$$

Scenario I



$$I_k^{\text{TCP}} = \{I_k^{\text{UDP}}, \alpha_0^{k-1}\}$$
$$u_k = \mu_k(I_k^{\text{TCP}})$$

Mathematical Formulation

$$x_{k+1} = f(x_k, v_k, w_k), \quad k = 0, 1, \dots$$

$$v_k = \alpha_k u_k \quad \text{or} \quad v_k = v_{k-1} \quad \text{if} \quad \alpha_k = 0$$

$$y_k = \beta_k z_k \quad z_k = h(x_k, w_k)$$

$\{\alpha_k\}, \{\beta_k\}$ independent *i.i.d.* Bernoulli

$$\text{Prob}(\alpha_k = 0) = \alpha, \quad \text{Prob}(\beta_k = 0) = \beta$$

$\{w_k\}$ *i.i.d.* plant / channel noise

Mathematical Formulation

$$\mathbf{x}_{k+1} = \mathbf{f}(\mathbf{x}_k, \mathbf{v}_k, \mathbf{w}_k), \quad k = 0, 1, \dots$$

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Control : $\mathbf{u}_k = \boldsymbol{\mu}_k (\mathbf{I}_k^{\text{TCP}})$
or $\mathbf{u}_k = \boldsymbol{\mu}_k (\mathbf{I}_k^{\text{UDP}})$

Mathematical Formulation

$$\mathbf{x}_{k+1} = f(\mathbf{x}_k, v_k, \mathbf{w}_k), \quad k = 0, 1, \dots$$

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$\{\alpha_k\}, \{\beta_k\}$ independent *i.i.d.* Bernoulli

$$\text{Prob}(\alpha_k = 0) = \alpha, \quad \text{Prob}(\beta_k = 0) = \beta$$

$$\text{PI} : E_{\mu} \left\{ q(\mathbf{x}_N) + \sum_k g(\mathbf{x}_k, v_k) \right\} =: \mathbf{J}(\mu_0^N, N)$$

$$\text{or } \limsup_{N \rightarrow \infty} (1/N) \mathbf{J}(\mu_0^N, N)$$

LQG with erasure channels

$$\mathbf{x}_{k+1} = \mathbf{A}\mathbf{x}_k + \alpha_k \mathbf{B}\mathbf{u}_k + \mathbf{w}_k, \quad k = 0, 1, \dots$$

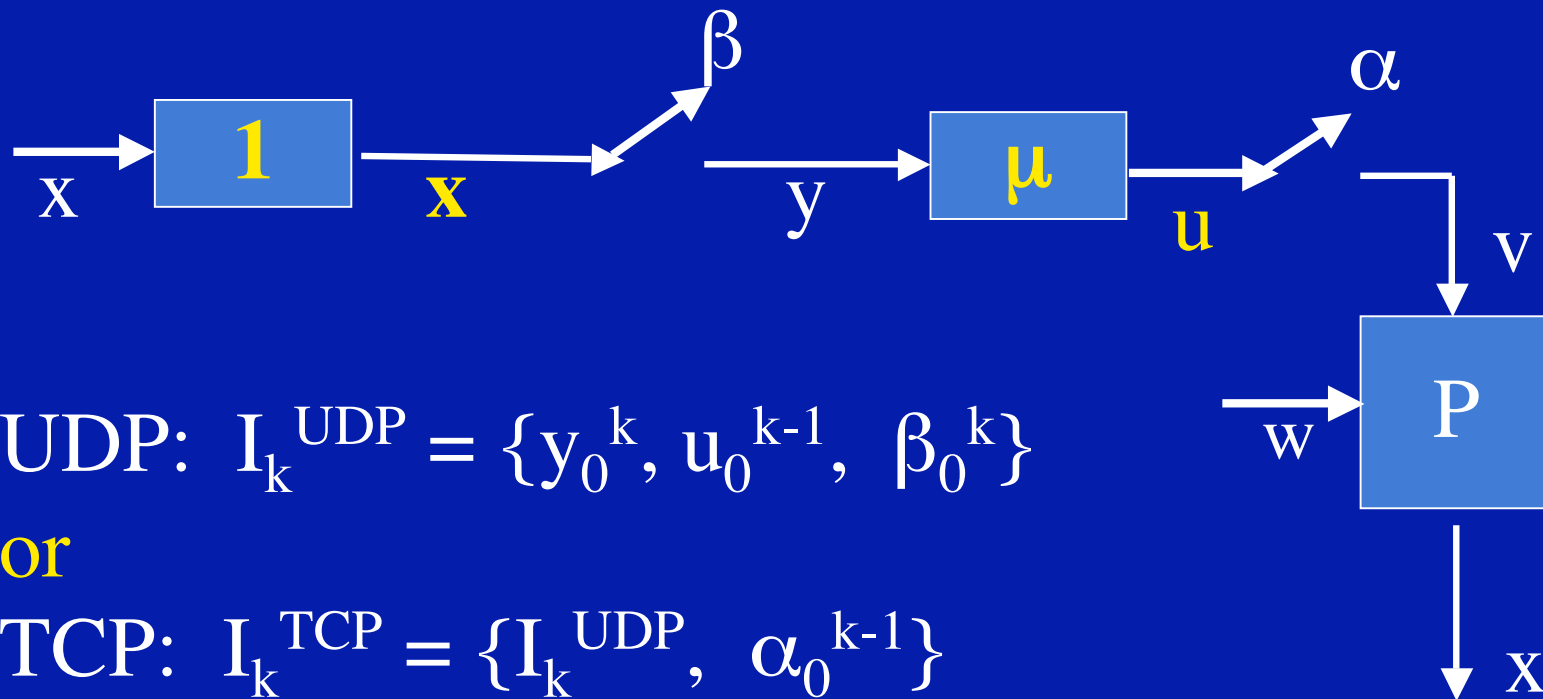
$$y_k = \beta_k x_k \quad z_k = x_k$$

$$\text{Prob}(\alpha_k = 0) = \alpha, \quad \text{Prob}(\beta_k = 0) = \beta$$

$$\mathbf{J}(\boldsymbol{\mu}_0^N, \mathbf{N}) = \mathbb{E}_{\boldsymbol{\mu}} \left\{ |\mathbf{x}_N|_F^2 + \sum_k |\mathbf{x}_k|_Q^2 + \alpha_k |\mathbf{u}_k|_R^2 \right\}$$

$$\text{or } \limsup_{N \rightarrow \infty} (1/N) \mathbf{J}(\boldsymbol{\mu}_0^N, \mathbf{N})$$

LQG with erasure: Two Scenarios



$$\text{UDP: } \mathbf{I}_k^{\text{UDP}} = \{y_0^k, u_0^{k-1}, \beta_0^k\}$$

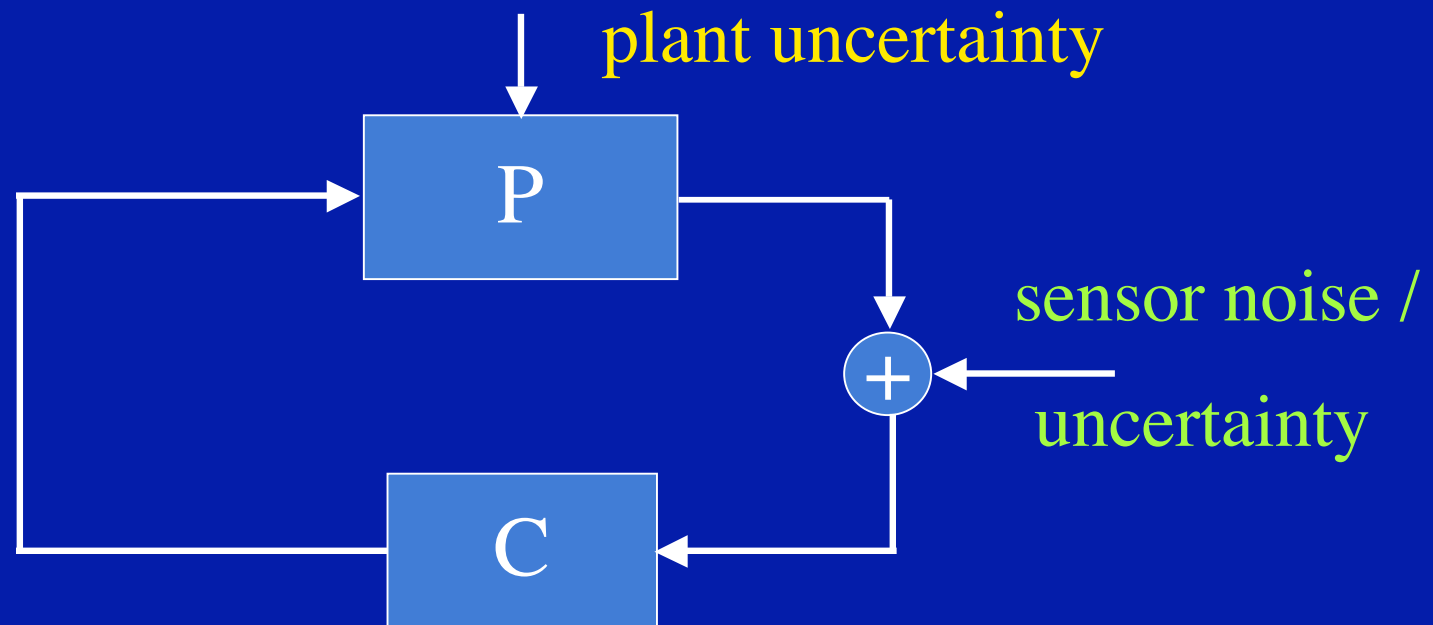
or

$$\text{TCP: } \mathbf{I}_k^{\text{TCP}} = \{\mathbf{I}_k^{\text{UDP}}, \alpha_0^{k-1}\}$$

$$u_k = \mu_k(\mathbf{I}_k)$$

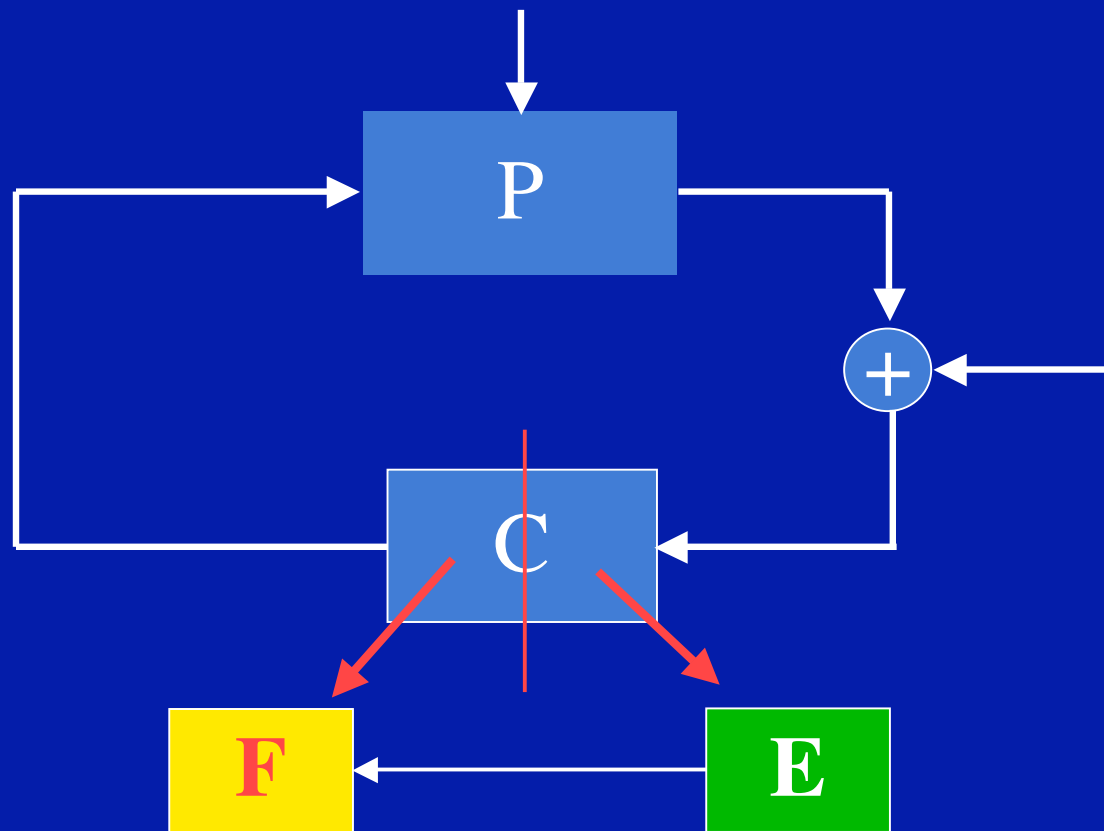
$$\text{PI} = E_{\mu} \left\{ |x_N|_F^2 + \sum |x_k|_Q^2 + \alpha_k |u_k|_R^2 \right\}$$

Digression: General Controller Design



dual role of control: action & probing
generally not aligned

Standard LQG: **Separation / Neutrality**



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Back to LQG with Erasure
With TCP: Finite Horizon

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With TCP: Finite Horizon

Separation of estimation and control
no dual effect

With TCP: Finite Horizon

- Separation of estimation and control; no dual effect
- $u_k = G_k E[x_k | I_k]$
 $= - (R + B^T K_{k+1} B)^{-1} B^T K_{k+1} A E[x_k | I_k]$

$$E[x_k | I_k] = A E[x_{k-1} | I_{k-1}] + \alpha_{k-1} B u_{k-1} \quad \text{if } \beta_k = 0$$
$$= x_k \quad \text{if } \beta_k = 1$$

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If noisy measurements, replace with
KF with intermittent measurements

With TCP: Finite Horizon

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$$K_k = \text{Ric}_{\text{TCP}}(K_{k+1}) \quad K_N = F$$

$$\text{Ric}_{\text{TCP}}(K) = A^T K A + Q - (1 - \alpha) A^T K B (R + B^T K B)^{-1} B^T K A$$

With TCP: Infinite Horizon

$$\begin{aligned} u_k &= G E[x_k | I_k] \\ &= - (R+B^T K B)^{-1} B^T K A E[x_k | I_k] \end{aligned}$$

$$K \leftarrow \text{Ric}_{\text{TCP}}(K) \Rightarrow K = \text{Ric}_{\text{TCP}}(K)$$

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If (A, Q) is observable, unique limit $K > 0$ iff
 $M \leftarrow A^T M A - (1 - \alpha) A^T M B (B^T M B)^{-1} B^T M A$
converges from $M_0 = I$ (based on Koning'82)

With TCP: Infinite Horizon

$$\begin{aligned} u_k &= G E[x_k | I_k] \\ &= - (R + B^T K B)^{-1} B^T K A E[x_k | I_k] \end{aligned}$$

If B is square and invertible, condition reduces to
 $|\lambda(A)| < 1 / \sqrt{\alpha}$

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$$\begin{aligned} u_k &= G E[x_k | I_k] \\ &= - (R + B^T K B)^{-1} B^T K A E[x_k | I_k] \end{aligned}$$

Control is m.s. stabilizing if, in addition,
 $|\lambda(A)| < 1 / \sqrt{\beta}$

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With UDP: Finite Horizon

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No separation of estimation and control
Dual effect!

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But, estimator can still be designed separately:

$$u_k = G_k(\alpha, \beta) E[x_k | I_k]$$

$$\begin{aligned} E[x_k | I_k] &= A E[x_{k-1} | I_{k-1}] + (1-\alpha) B u_{k-1} && \text{if } \beta_k = 0 \\ &= x_k && \text{if } \beta_k = 1 \end{aligned}$$

With UDP: Finite Horizon

No separation of estimation and control
Dual effect!

But, estimator can still be designed separately:

$$u_k = G_k(\alpha, \beta) E[x_k | I_k]$$

And G involves two recursions ($P_N = 0, K_N = F$):

$$P_k = \beta A^T P_{k+1} A - (1 - \alpha) A^T K_{k+1} B G_k$$

$$K_k = A^T K_{k+1} A - P_{k+1} + \beta A^T P_{k+1} A + Q$$

$$G_k = - (R + B^T (K_{k+1} + \alpha \beta P_{k+1}) B)^{-1} B^T K_{k+1} A$$

With UDP: Infinite Horizon

$$P \leftarrow \beta A^T P A - (1 - \alpha) A^T K B G$$

$$K \leftarrow A^T K A - P + \beta A^T P A + Q$$

$$G = - (R + B^T (K + \alpha \beta P) B)^{-1} B^T K A$$

With UDP: Infinite Horizon

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Simpler test, equivalent condition if (A, Q) observable
 ($Q = R = 0, F = I$ in the above)

$$\begin{aligned} \Pi \leftarrow & (1 - \alpha) A^T \Lambda B (B^T (\Lambda + \alpha \beta \Pi) B)^{-1} B^T \Lambda A \\ & + \beta A^T \Pi A \quad \Pi_0 = 0 \end{aligned}$$

$$\begin{aligned} \Lambda \leftarrow & -(1 - \alpha) A^T \Lambda B (B^T (\Lambda + \alpha \beta \Pi) B)^{-1} B^T \Lambda A \\ & + A^T \Lambda A \quad \Lambda_0 = I \end{aligned}$$

With UDP: Infinite Horizon

$$P \leftarrow \beta A^T P A - (1 - \alpha) A^T K B G$$

$$K \leftarrow A^T K A - P + \beta A^T P A + Q$$

$$G = - (R + B^T (K + \alpha \beta P) B)^{-1} B^T K A$$

If Π and Λ converge, so do P and K , and G is the optimal steady-state gain

$$\begin{aligned} \Pi \leftarrow & (1 - \alpha) A^T \Lambda B (B^T (\Lambda + \alpha \beta \Pi) B)^{-1} B^T \Lambda A \\ & + \beta A^T \Pi A \quad \Pi_0 = 0 \end{aligned}$$

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With UDP: Infinite Horizon

$$P = \beta A^T P A - (1 - \alpha) A^T K B G$$

$$K = A^T K A - P + \beta A^T P A + Q$$

$$G = - (R + B^T (K + \alpha \beta P) B)^{-1} B^T K A$$

Then, the steady-state optimal control is

$$u_k = G(\alpha, \beta) E[x_k | I_k]$$

$$\begin{aligned} \Pi \leftarrow & (1 - \alpha) A^T \Lambda B (B^T (\Lambda + \alpha \beta \Pi) B)^{-1} B^T \Lambda A \\ & + \beta A^T \Pi A \quad \Pi_0 = 0 \end{aligned}$$

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Extensions

- Actuator applies last available control (instead of “zero”) when the control packet is lost.

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- Noisy measurements
- Correlated link failure processes
- With multiple channels, packets being lost with different probabilities on each sensor and actuator link
- Delays in the acknowledgements with TCP

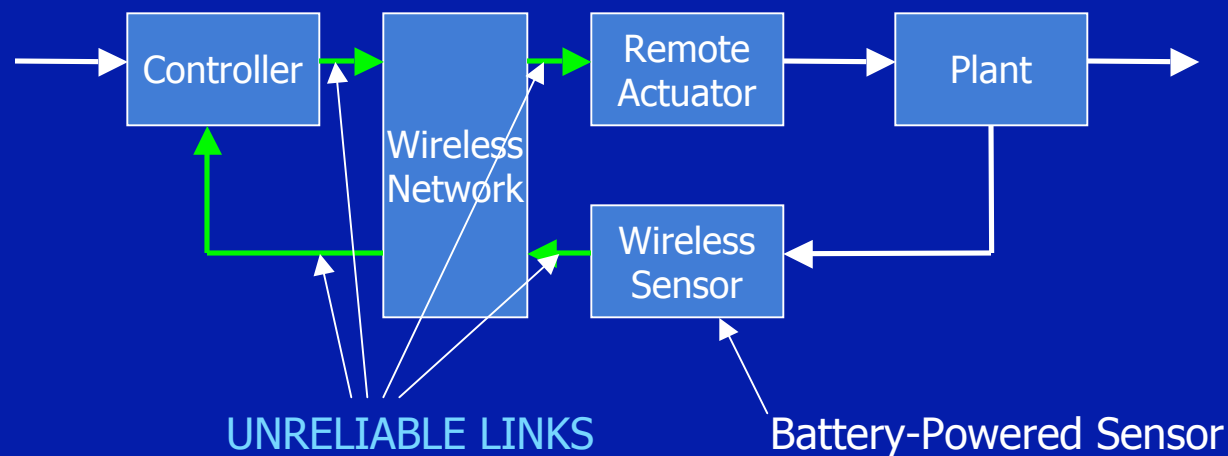
Other Set-ups

Usage-Constrained Estimation and Control

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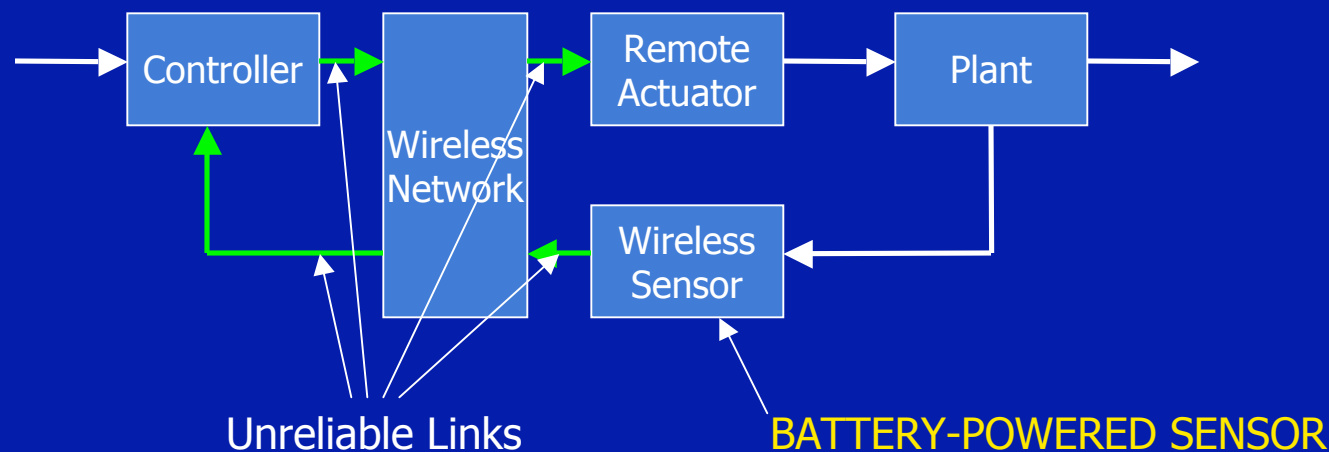
Wireless Networks & Control

- Estimation & Control when **controller-plant communication** takes place over a **Wireless Network**



Wireless Networks & Control

- Estimation & Control when **controller-plant communication** takes place over a **Wireless Network**



Limited Usage in Sensing & Control

Reasons

- To conserve battery power in wireless sensors
(with RF channels a significant amount of power is consumed to transmit sensor measurements)
- Sharing of bandwidth among several users
(time allocation limited transmission -- TDM systems)
- Power limitation on controller
- Limits on the frequency of interaction with the plant/system
- CANs -- actuator, controller, sensor connected over a serial bus -- access one at a time

Limited Usage in Sensing & Control

Reasons

- To conserve battery power in wireless sensors (with RF channels a significant amount of power is consumed to transmit sensor measurements)
- Sharing of channels with other sensors (time allocation)
- Power consumption
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- CANs -- actuator, controller, sensor connected over a serial bus -- access one at a time

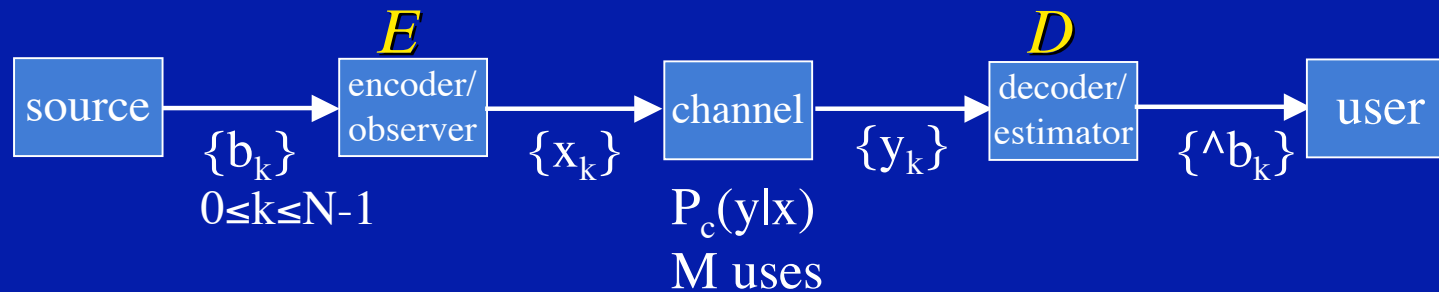
**Sensing and Control
are Expensive !**

Three Classes of Problems

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PROBLEM CLASS I

Optimal Estimation over a Limited-Use Channel



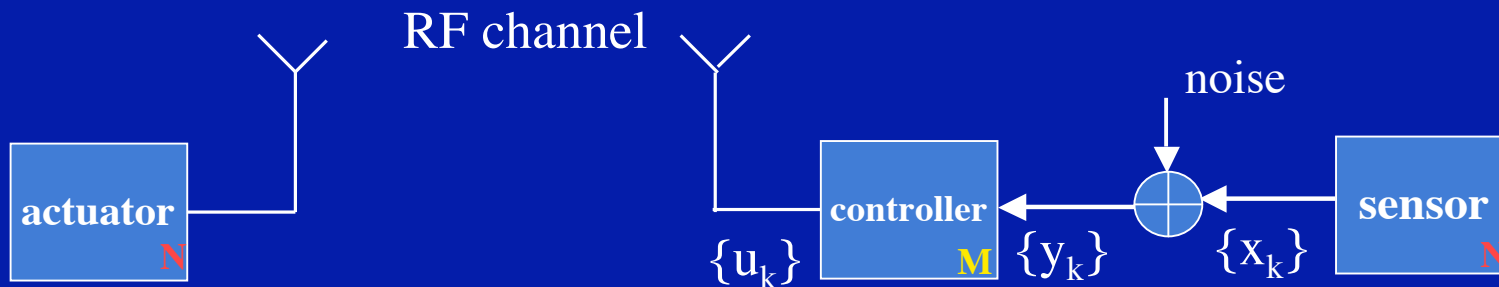
$$x_k = E(z_k) \quad x \in X \quad y \in Y$$

$$z_k = b_k + v_k \quad M < N$$

Given a “source” and a “memoryless channel”, for a given message length N , and number of channel uses M , what is the minimum attainable value of the average distortion $D_{(M,N)}$ and a corresponding E & D pair?

PROBLEM CLASS II

Control over a Limited-Use Channel



$$x_{k+1} = f(x_k, u_k, w_k) \quad x_k \in X_k, u_k \in U_k, y \in Y_k$$

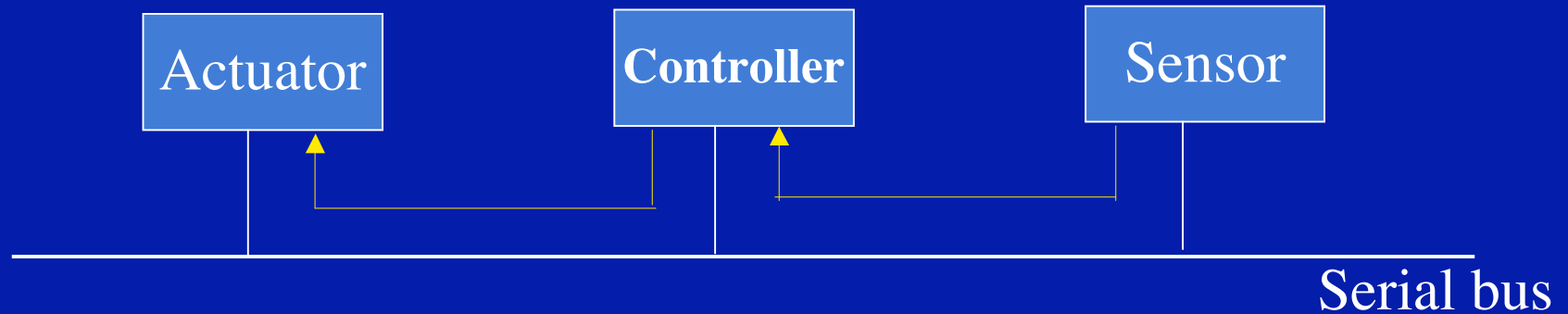
$$y_k = h_k(x_k) + v_k \quad M < N$$

$$u_k = \mu_k(I_k), I_k := \{y_{[0,k]}, u_{[0,k-1]}\}, \mu_k \rightarrow U_k \text{ only } M \text{ times}$$

Given a horizon of N units, and with controller allowed to transmit for only $M < N$ times, what is the minimum attainable value of a performance index J , and a corresponding controller?

PROBLEM CLASS III

Scheduled Measurements & Controls



$$x_{k+1} = A x_k + a_k u_k + w_k, \quad k = 0, \dots, N-1$$

a_k is 0 or 1

$$y_k = (1 - a_k) x_k, \quad k = 0, \dots, N-1 \quad \text{measurement}$$

Controller cannot receive and send simultaneously

Solutions

In all cases, solutions are of the **threshold** type.

For example, for Problem Class I, best sensor policy is of the form:

At time k transmit z_k if it is in a measurable set $\mathbb{Y}(s_k, t_k)$, otherwise do not.

$\mathbb{Y}(s, t)$ obtained offline as the minimizer in a recursive equation satisfied by accumulated optimum error, $e^*(s, t)$, at each point (s, t) , where t is # decision instances left, and s is # usages left.

Specific structure of $\mathbb{Y}(s, t)$ depends on the pdf/pmf and PI.

Smart Sensing

Input parameters:

1. **Planning horizon:** Desired length of time the sensor will be in operation (**N**)
2. **Battery size:** Size of the power source installed into the sensor (**M**)
3. **Process model (data):** This is application specific, and can be gathered from a variety of sources including, measurement/historic data, modeling, etc. (**Source Statistics**)
4. **Performance criterion:** (**Distortion Measure**)

Smart Sensing

How?

- Given the input parameters and the channel compute the optimum transmission and decoding policies for the wireless sensor & the receiver
- The computation is “power aware” (M), and “planning-horizon aware” (N)
- Minimal online computational complexity (lookup table, binary comparisons,...)
- Offline computational complexity will depend on the source and channel models

To conclude ...

- Networks & Control
- Issues in Networked Control Systems
 - What/when/how to transmit & control
- Control with Unreliable Channels
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 - When to measure & when to control
- **Conclusions**

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- Networks & Control

Networked Control Systems is a current, active, and fertile research topic offering interesting challenges in both theory and applications.

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- Control with Power-Limited Communication

-- When to measure

- Conclusions

Thanks !