

# Where do stabilization problems sit in the communication problem hierarchy?

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April 1, 2005

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*April 1, 2005 at CalTech Workshop*

## Outline

1. Communication problem hierarchy
2. Stabilization problems and anytime communication
3. Core reason: feedback channel coding: block-length is *not* a good proxy for delay

## Big questions about communication.

- Are all communication problems asymptotically alike?

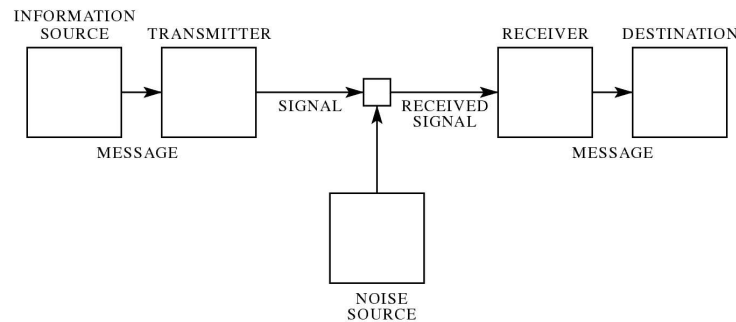
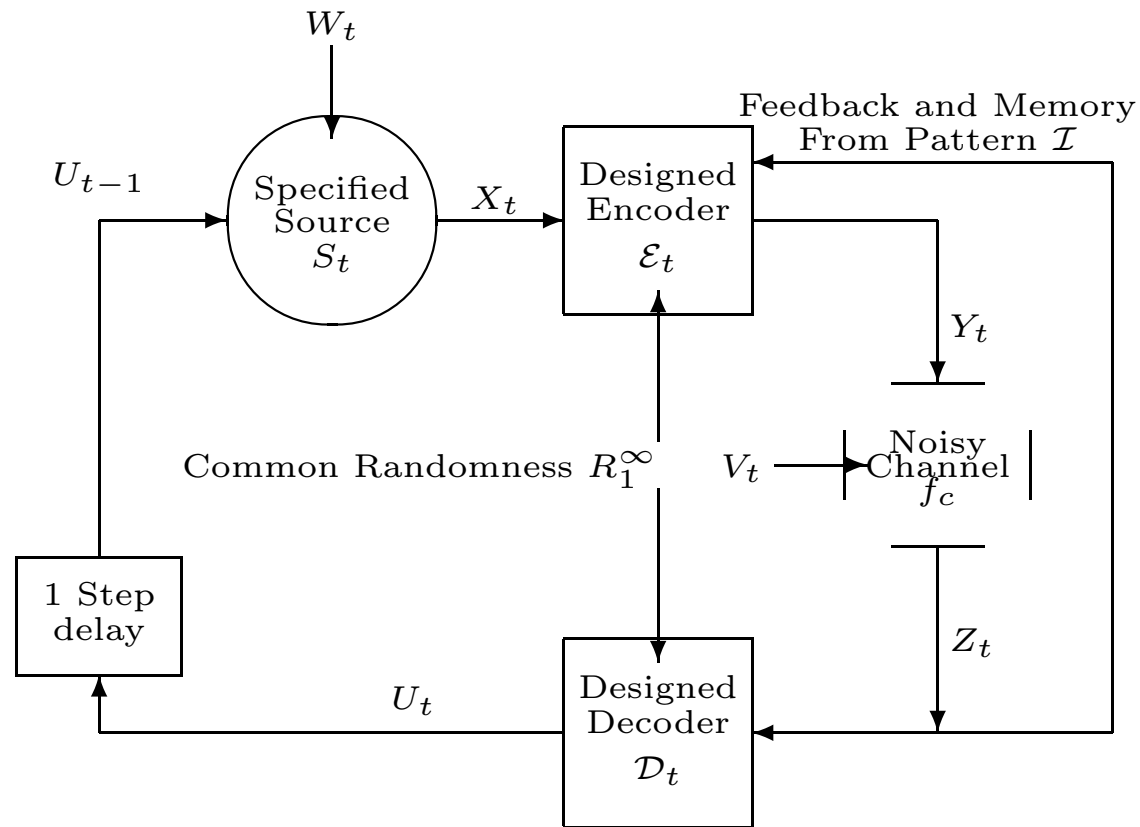


Fig. 1—Schematic diagram of a general communication system.

- How does delay interact with capacity issues?
- Can we find examples that let us explore these questions in an asymptotic setting?

“... can be pursued further and is related to a duality between past and future and the notions of control and knowledge. Thus we may have knowledge of the past and cannot control it; we may control the future but have no knowledge of it.” — Claude Shannon 1959

## An abstract model of single channel problems



- Problem: Source  $S$ , Information pattern  $\mathcal{I}$ , and Objective  $\mathcal{V}$ .
- Constrained resource: Noisy channel  $f_c$
- Designed solution: “Encoder”  $\mathcal{E}$ , “Decoder”  $\mathcal{D}$

## Focus: what channels are “good enough” for the problem

- $f_c$  solves the problem if  $\exists \mathcal{E}, \mathcal{D}$  so system satisfies  $\mathcal{V}$
- Problem  $A$  is *harder* than problem  $B$  if any  $f_c$  that solves  $A$  solves  $B$ .
- Information theory is an *asymptotic* theory
  - Pick  $\mathcal{V}$  family with an appropriate “slack” parameter
  - Consider the set of channels that solve the problem.
  - Take union over slack parameter choices.

## The Shannon problems $A_{R,\epsilon,d}$

- Source: *noninteractive*  $X_i$  ( $R$  bits): fair coin tosses
- Information pattern:  $\mathcal{D}_i$  has access to  $Z_1^i$ 
  - $A^f$  With feedback:  $\mathcal{E}_i$  gets  $X_1^i$  and  $Z_1^{i-1}$
  - $A^{nf}$  Without feedback:  $\mathcal{E}_i$  gets only  $X_1^i$
- Performance objective:  $\mathcal{V}(\epsilon, d)$  is satisfied if  $\mathcal{P}(X_i \neq U_{i+d}) \leq \epsilon$  for every  $i \geq 0$ .
  - Slack parameter: permitted delay  $d$
  - Natural orderings: larger  $\epsilon, d$  is easier but larger  $R$  is harder.
- Classical capacity

$$\mathcal{C}_R^f = \bigcap_{\epsilon > 0} \bigcap_{R' < R} \bigcup_{d > 0} \{f_c | f_c \text{ solves } A_{R',\epsilon,d}^f\}$$

$$C_{\text{Shannon}}(f_c) = \sup\{R > 0 | f_c \in \mathcal{C}_R\}$$

## Classical relationships

- Feedback doesn't change capacity for memoryless channels  $\mathcal{C}^m$

$$\mathcal{C}_R^{nf} \cap \mathcal{C}^m = \mathcal{C}_R^f \cap \mathcal{C}^m$$

- Zero-error capacity

$$\mathcal{C}_{0,R}^f = \bigcap_{R' < R} \bigcup_{d > 0} \{f_c | f_c \text{ solves } A_{R',0,d}^f\}$$

$$C_0(f_c) = \sup\{R > 0 | f_c \in \mathcal{C}_{0,R}\}$$

- Can change with feedback even for memoryless channels

$$(\mathcal{C}_{0,R}^f \cap \mathcal{C}^m) \subset (\mathcal{C}_{0,R}^f \cap \mathcal{C}^m)$$

- Zero-error problem is fundamentally harder

$$(\mathcal{C}_{0,R}^{nf} \cap \mathcal{C}^m) \subset (\mathcal{C}_{0,R}^f \cap \mathcal{C}^m) \subset (\mathcal{C}_R \cap \mathcal{C}^m)$$

## Estimation with distortion: $A_{(F_X, \rho, D, d)}$

- Source: *noninteractive*  $X_i$  drawn iid from  $F_X$
- Same information patterns: with/without feedback.
- Performance objective:  $\mathcal{V}(\rho, D, d)$  is satisfied if  $\lim_{n \rightarrow \infty} \frac{1}{n} E[\sum_{i=1}^n \rho(X_i, U_{i+d})] \leq D$ .
  - Slack parameter: permitted delay  $d$
  - Natural orderings: larger  $D, d$  is easier
- Channels that are good enough

$$\mathcal{C}_{e, (F_X, \rho, D)}^f = \bigcap_{D' > D} \bigcup_{d > 0} \{f_c | f_c \text{ solves } A_{(F_X, \rho, D', d)}^f\}$$

- “Separation Theorem” if  $\rho$  is finite.

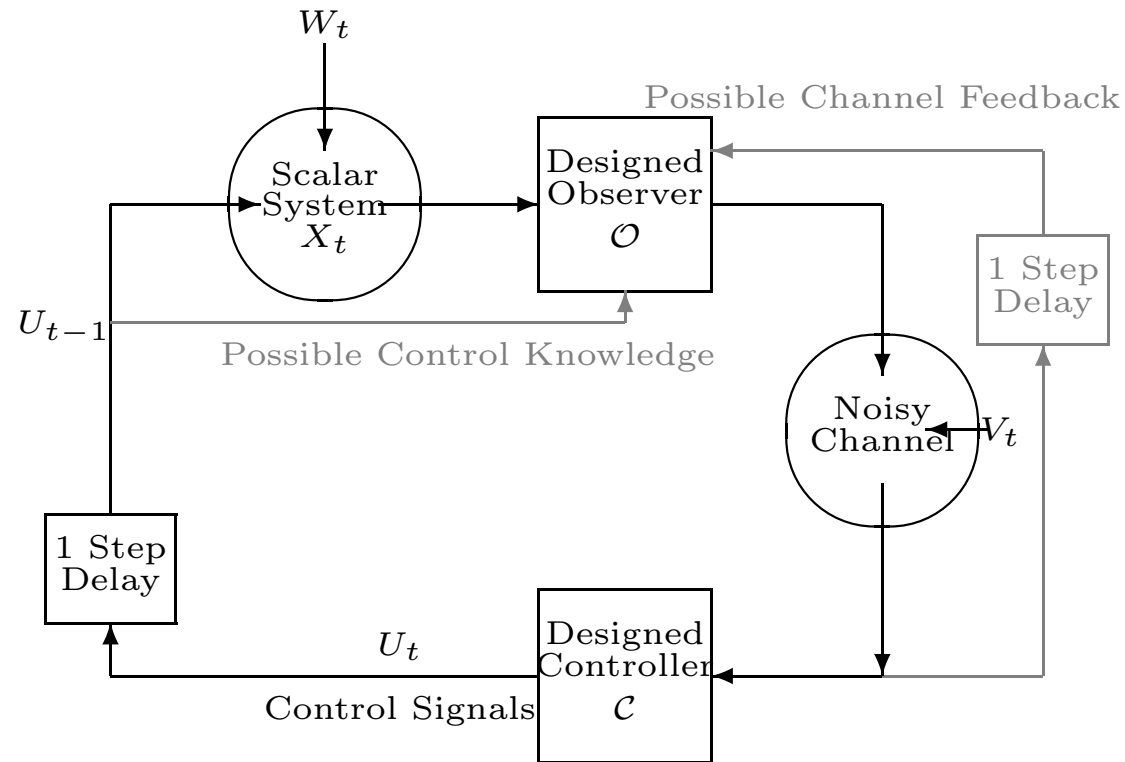
$$(\mathcal{C}_{R(D)} \cap \mathcal{C}^m) = (\mathcal{C}_{e, (F_X, \rho, D)}^{nf} \cap \mathcal{C}^m) = (\mathcal{C}_{e, (F_X, \rho, D)}^f \cap \mathcal{C}^m)$$



## Stabilization problems and anytime communication

- Simple control problem
- Why classical capacity is not enough.
- Why anytime (delay-universality) is needed
- Implications (power laws, etc.)
- Imperfect information patterns and implicit communication

## Our simple scalar distributed control problem



$$X_{t+1} = \lambda X_t + U_t + W_t$$

- Unstable  $\lambda > 1$ , bounded initial condition and disturbance  $W$ .
- Goal: Stability =  $\sup_{t>0} E[|X_t|^\eta] \leq K$  for some  $K < \infty$ .

## Is Shannon capacity all we need?

- Consider a system with
  - $\lambda = 2$  for the dynamics
  - noisy channel that sometimes drops packets but is otherwise noiseless (Real erasure channel)

$$Z_t = \begin{cases} Y_t & \text{with Probability } \frac{1}{2} \\ 0 & \text{with Probability } \frac{1}{2} \end{cases}$$

- No other constraints, so design is obvious:  $Y_t = X_t$  and  $U_t = -\lambda Z_t$
- Resulting closed loop dynamics:

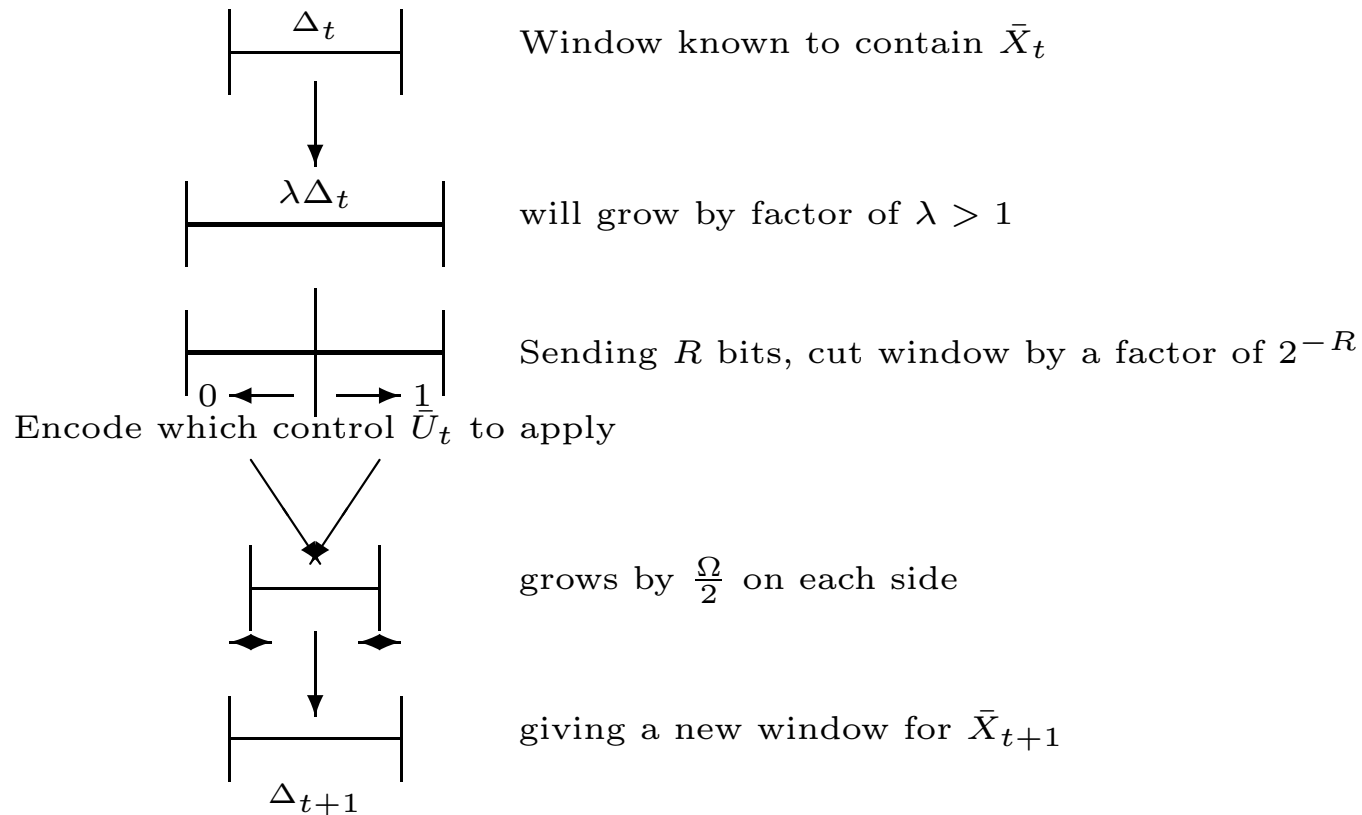
$$X_{t+1} = \begin{cases} W_t & \text{with Probability } \frac{1}{2} \\ 2X_t + W_t & \text{with Probability } \frac{1}{2} \end{cases}$$

## Is the closed-loop system stable?

$$X_{t+1} = \begin{cases} W_t & \text{with Probability } \frac{1}{2} \\ 2X_t + W_t & \text{with Probability } \frac{1}{2} \end{cases}$$

- i.i.d. erasures mean arbitrarily long stretches of erasures are possible, though unlikely.
  - System is not guaranteed to stay inside any box.
  - Under stochastic disturbances, the variance of the state is asymptotically infinite.
- For worst case disturbances  $W_t = 1$ , the tail probability is dying off as  $P(|X| > x) \approx \frac{K}{x}$ .
- Meanwhile,  $C = \infty$ !

## Run same plant $\bar{X}$ over noiseless channel



**As long as  $R > \log_2 \lambda$ , we can have  $\Delta$  stay bounded forever.**

## What is needed: key intuition

- Break state  $X$  into sum of  $\check{X}$  (response to disturbance) and  $\tilde{X}$  (response to control)
- Suppose  $\lambda = 2$  and so  $\check{X}_t = \sum_{i=0}^t 2^i W_{t-1}$
- Assume  $W_j$  either 0 or 1
- In binary notation:  $\check{X}_t = W_0 W_1 W_2 \dots W_{t-1}.00000\dots$
- If  $-\tilde{X}_t$  is close to  $\check{X}_t$ , their binary representations likely agree in all the high-order bits.
  - High-order bits represent earlier disturbances.
  - Typically, to get a difference at the  $W_{t-d}$  level, we have to be off by about  $2^d$ .

**Stabilization implies communicating bits reliably in an anytime fashion.**

## Anytime communication problems: $A_{R,\alpha,K}$

- Same as Shannon problem in source and information pattern.
- Performance objective different:
  - Reinterpret  $U_t = 0.\hat{X}_0(t), \hat{X}_1(t), \hat{X}_2(t), \dots$  in binary
  - $\mathcal{V}_{(K,\alpha)}$  is satisfied if  $\mathcal{P}(X_i \neq \hat{X}_i(i+d)) \leq K2^{-\alpha d}$  for every  $i \geq 0, d \geq 0$ .
  - Slack parameter: constant factor  $K$
  - Natural orderings: larger  $K$  is easier, but larger  $R, \alpha$  are harder.
- Capacity

$$\mathcal{C}_{a,(R,\alpha)}^f = \bigcap_{R' < R} \bigcap_{\alpha' < \alpha} \bigcup_{K > 0} \{f_c | f_c \text{ solves } A_{(R',\alpha',K)}^f\}$$

$$C_{\text{any}}(f_c, \alpha) = \sup\{R > 0 | f_c \in \mathcal{C}_{a,(R,\alpha)}^f\}$$

## Separation theorem for control

*Necessity:* If a scalar system with parameter  $\lambda > 1$  can be stabilized with finite  $\eta$ -moment across a noisy channel, then the **channel with noiseless feedback** must have

$$C_{\text{any}}(\eta \log_2 \lambda) \geq \log_2 \lambda$$

In general: If  $P(|X| > m) < f(m)$ , then  $\exists K : P_{\text{error}}(d) < f(K \lambda^d)$

*Sufficiency:* If there is an  $\alpha > \eta \log_2 \lambda$  for which the **channel with noiseless feedback** has

$$C_{\text{any}}(\alpha) > \log_2 \lambda$$

then the scalar system with parameter  $\lambda \geq 1$  with a bounded disturbance can be stabilized across the noisy channel with finite  $\eta$ -moment **assuming nested information.**



## What does all this imply?

- If we want  $P(|X_t| > m) \leq f(m) = 0$  for some finite  $m$ , we require zero-error reliability across the channel.
- For generic DMCs, anytime reliability with feedback is upper-bounded:

$$\begin{aligned} f(K\lambda^d) &\geq \zeta^d \\ f(m) &\geq \zeta^{\frac{\log_2(\frac{m}{K})}{\log_2 \lambda}} \\ f(m) &\geq K' m^{-\frac{\log_2 \frac{1}{\zeta}}{\log_2 \lambda}} \end{aligned}$$

**A controlled state can have at best a power-law tail.**

- If we just want  $\lim_{m \rightarrow \infty} f(m) = 0$ , then just Shannon capacity  $> \log_2 \lambda$  is required for DMCs.
- Almost-sure stabilization for  $W_t = 0$  follows by time-varying transformation.

## Known feedback anytime capacities

Characterizing the boundary of possible  $(R, \alpha)$  pairs. For generic DMCs, have both upper and lower bounds.

- $L$ -bit packet erasure channel

$$C_{\text{any}}(\alpha) = \frac{\alpha L}{\alpha + \log_2\left(\frac{1-\delta}{1-\delta 2^\alpha}\right)}$$

- Variable-sized packet erasure channel with expected packet-size constrained to be  $\bar{L}$  and maximum packet-size  $L_{max}$  (Allerton 2004)

$$C_{\text{any}}(\alpha) = \min \left( (1 - \delta)\bar{L}, \frac{\alpha L_{max}}{\alpha + \log_2\left(\frac{1-\delta}{1-\delta 2^\alpha}\right)} \right)$$

- Average Power-constrained AWGN or Gilbert-Elliott with CSI (ISIT 05)

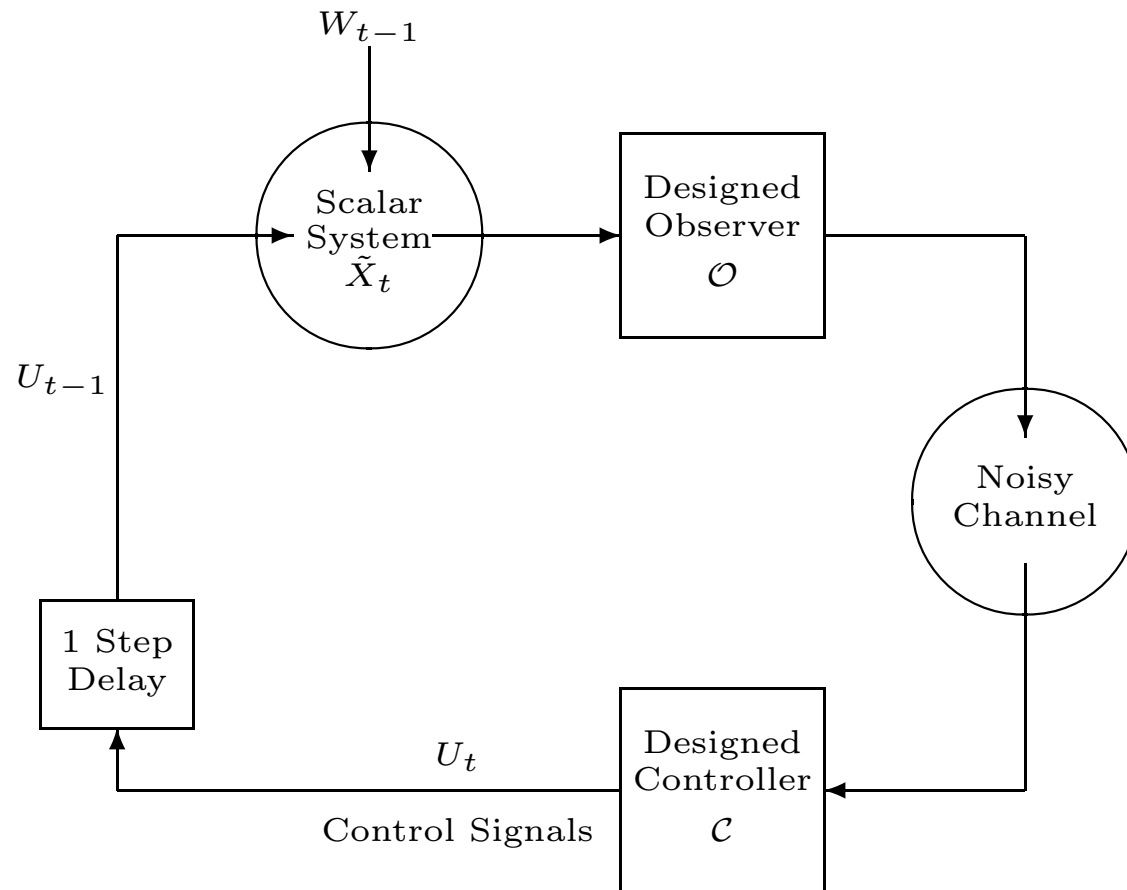
$$C_{\text{any}}(\alpha) = \frac{1}{2} \log_2 \left( 1 + \frac{P}{\sigma^2} \right) = C_{\text{Shannon}}$$

- Power-constrained AWGN+erasure (Allerton 2004)

$$\alpha^*(R) = \begin{cases} -\log_2 \delta & \text{if } R < C_{\text{Shannon}} \\ 0 & \text{otherwise} \end{cases}$$

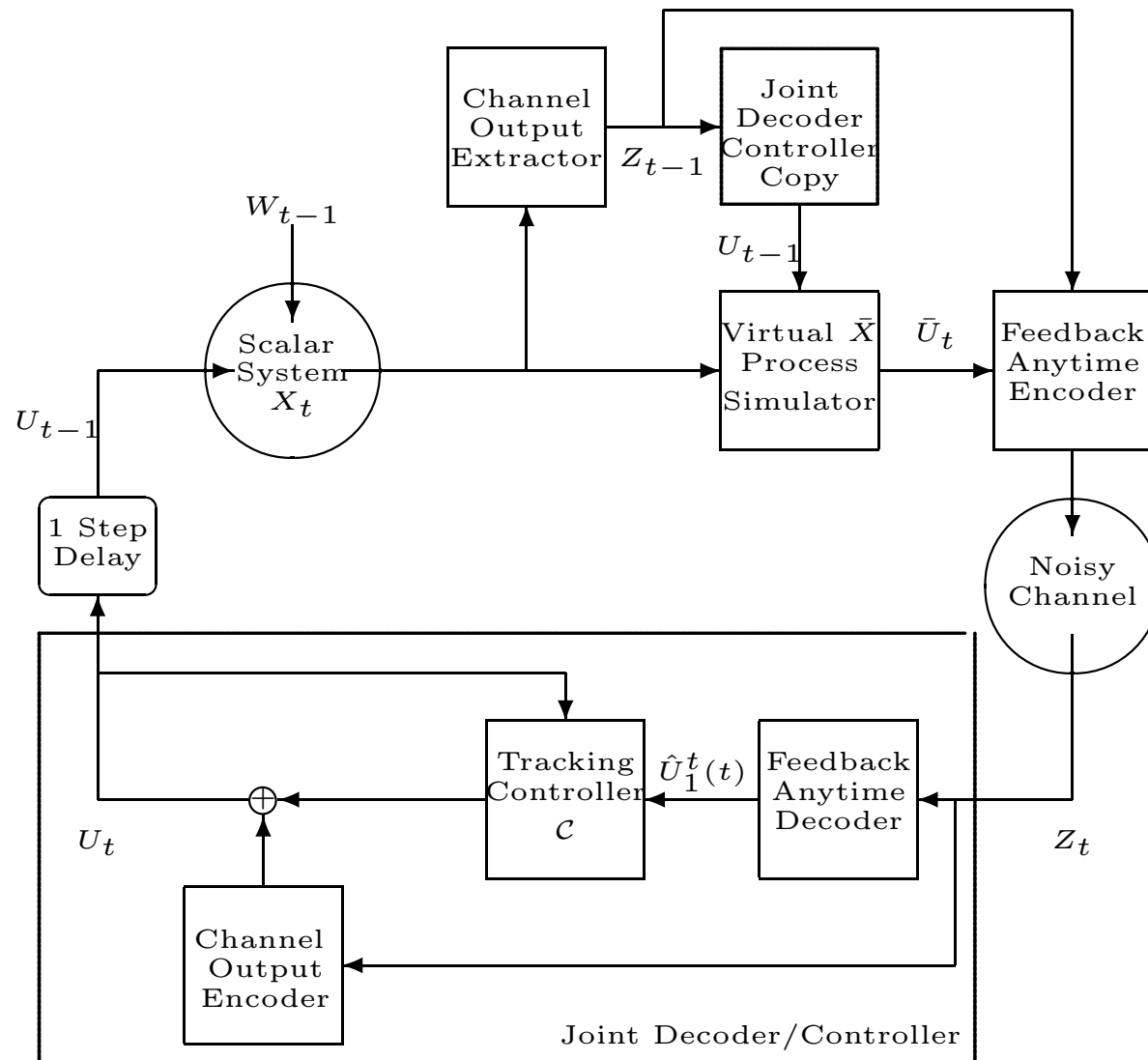
- Similar bound for Markov channels with CSI, but replace  $\delta$  with the largest eigenvalue of a censored transition matrix.

## What about imperfect information patterns?



- Do we now need a higher quality channel?
- The only path from the controller to the observer is through the plant.

Make the plant “dance” in a stable way!



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## Stabilization and Anytime Equivalence

- With nested information:  $A_{\lambda,\eta,K}^f$ . Without:  $A^{nf}$ 
  - Slack parameter:  $K$  (Performance)
  - Natural ordering: larger  $\eta, \lambda$  are harder, but larger  $K$  is easier.

$$\mathcal{C}_{s,(\lambda,\eta)}^f = \bigcap_{\lambda' < \lambda} \bigcap_{\eta' < \eta} \bigcup_{K > 0} \{f_c \mid f_c \text{ solves } A_{(\lambda',\eta',K)}^f\}$$

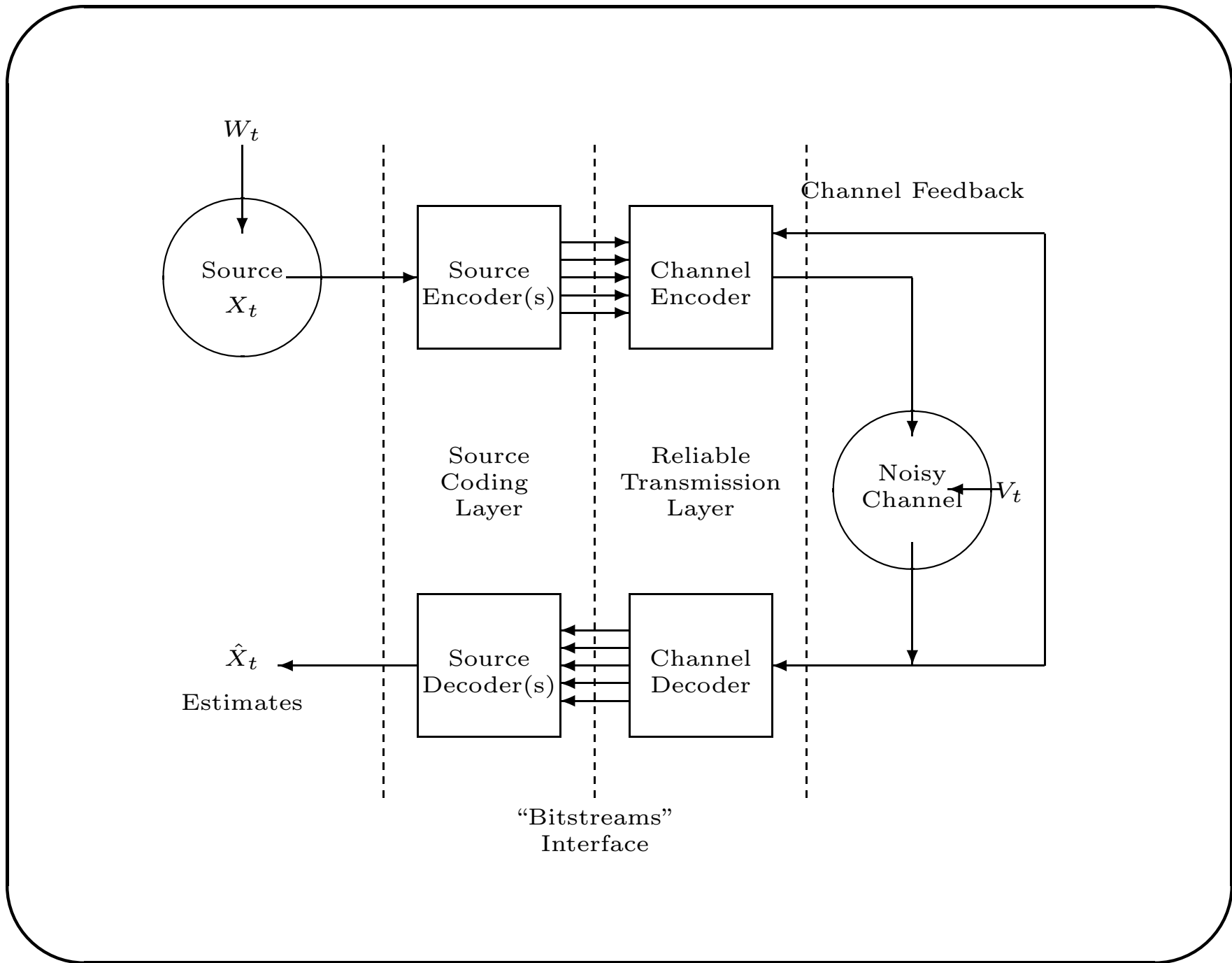
- Equivalences

$$\mathcal{C}_{s,(\lambda,\eta)}^{nf} \subseteq \mathcal{C}_{s,(\lambda,\eta)}^f = \mathcal{C}_{a,(\log_2 \lambda, \eta \log_2 \lambda)}^f$$

$$(\mathcal{C}_{s,(\lambda,\eta)}^{nf} \cap \mathcal{C}^{\text{finite}}) = (\mathcal{C}_{s,(\lambda,\eta)}^f \cap \mathcal{C}^{\text{finite}}) = (\mathcal{C}_{a,(\log_2 \lambda, \eta \log_2 \lambda)}^f \cap \mathcal{C}^{\text{finite}})$$

## The vector case: differentiated service

- Possibly many unstable eigenvalues
- All unstable eigenspaces need to be estimated with eventually zero error
- Some bits are more important than others.
- Instead of a single  $\alpha$  and a single rate  $R$ , we get a vector  $\vec{\alpha}$  and a rate vector  $\vec{R}$ .
- Direct and converse both hold on an eigenvalue by eigenvalue basis.
- Imperfect information patterns: need to deal with *intrinsic delays*.



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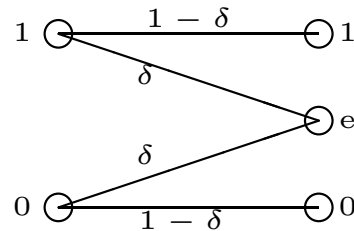
## *Asymptotic communication problem hierarchy*

- The easiest: Shannon communication
  - Asymptotically: a single figure of merit  $C$
  - Equivalent to most estimation problems of stationary ergodic processes with bounded distortion measures.
  - Feedback does not matter.
- Intermediate families: Anytime communication
  - Multiple figures of merit:  $(\vec{R}, \vec{\alpha})$
  - Feedback case equivalent to stabilization problems
  - Related nonstationary estimation problems fall here also
  - Feedback matters.
- Hardest level: Zero-error communication
  - Single figure of merit  $C_0$
  - Feedback matters.

## Outline

1. Communication problem hierarchy
2. Stabilization problems and anytime communication
3. Feedback channel coding: block-length is *not* a good proxy for delay

## Our favorite example: The BEC



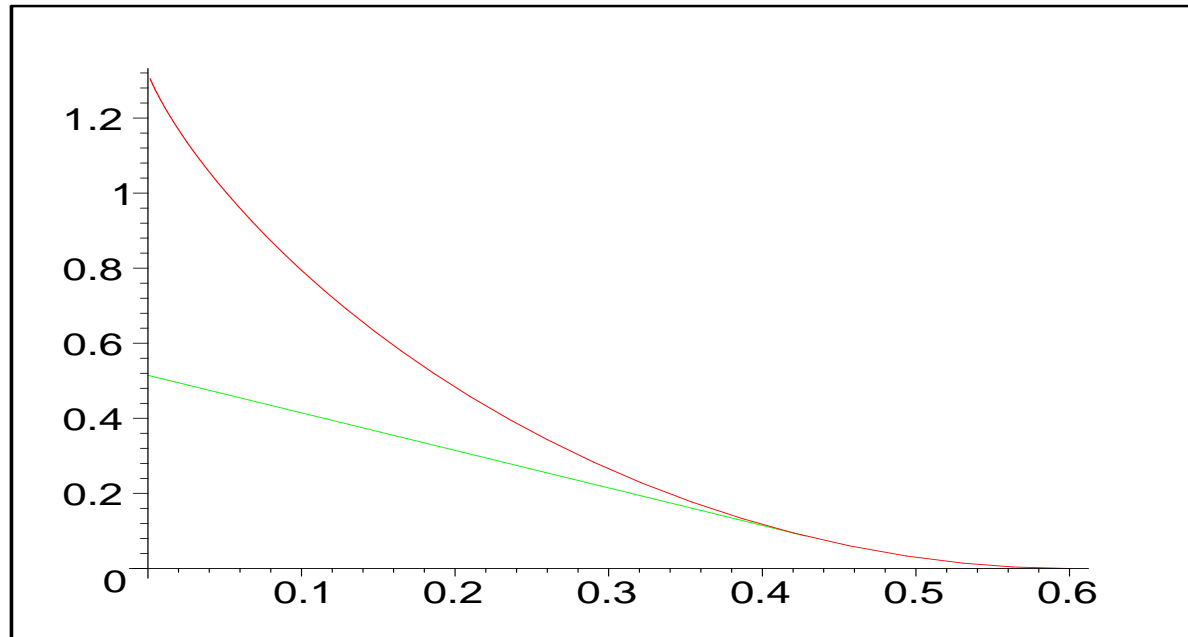
- Simple capacity  $1 - \delta$
- With perfect feedback, simple to achieve: retransmit until it gets through
  - Time till success:  $\text{Geometric}(1 - \delta)$
  - Expected time to get through:  $\frac{1}{1 - \delta}$
- One size fits all!
  - Strategy works regardless of  $\delta$  (Universality)
  - All bits eventually get through correctly
  - Gets bits through as soon as possible

## Is block-length a good proxy for delay?

- Study erasure case with and without feedback
- Block-codes vs general codes
- Behavior of probability of error with delay

## Fixed block length coding

$$E(R) = \lim_{n \rightarrow \infty} -\frac{\log_2 P_e(n)}{n}$$



- Classical bounds
  - Random coding bound  $\max_{\rho \in [0,1]} E_0(\rho) - \rho R$
  - Sphere-packing bound  $D(1 - R || \delta)$
- What happens with feedback?

## BEC reliability with feedback

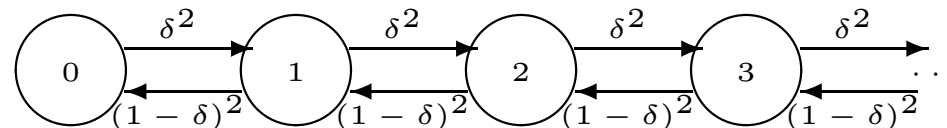
- At rate  $R < 1$ , have  $Rn$  bits to transmit in  $n$  channel uses.
- Typically  $(1 - \delta)n$  code bits will be received.
- Block errors caused by atypical channel behavior.
  - Doomed if fewer than  $Rn$  bits arrive intact.
  - *Feedback can not save us.*
  - $D(1 - R|\delta)$
- Dobrushin showed that this type of behavior is common.
  - For sufficiently symmetric channels, the sphere-packing bound  $E_{sp}(R)$  is unchanged with feedback.
  - In general, can get better but not by much — same convex  $\cup$  shape.

## Bit error vs Block error

- Block-code with feedback is unfair to later bits.
  - First bit:  $-\log_2 \delta$
  - Last bit:  $D(1 - R||\delta)$
- Symmetrize by randomly shuffling bits first.
- Makes no difference in exponential order!

## BEC with feedback

- $R = \frac{1}{2}$  example:



- Birth-death chain: positive recurrent if  $\delta < \frac{1}{2}$
- Delay exponent easy to see:

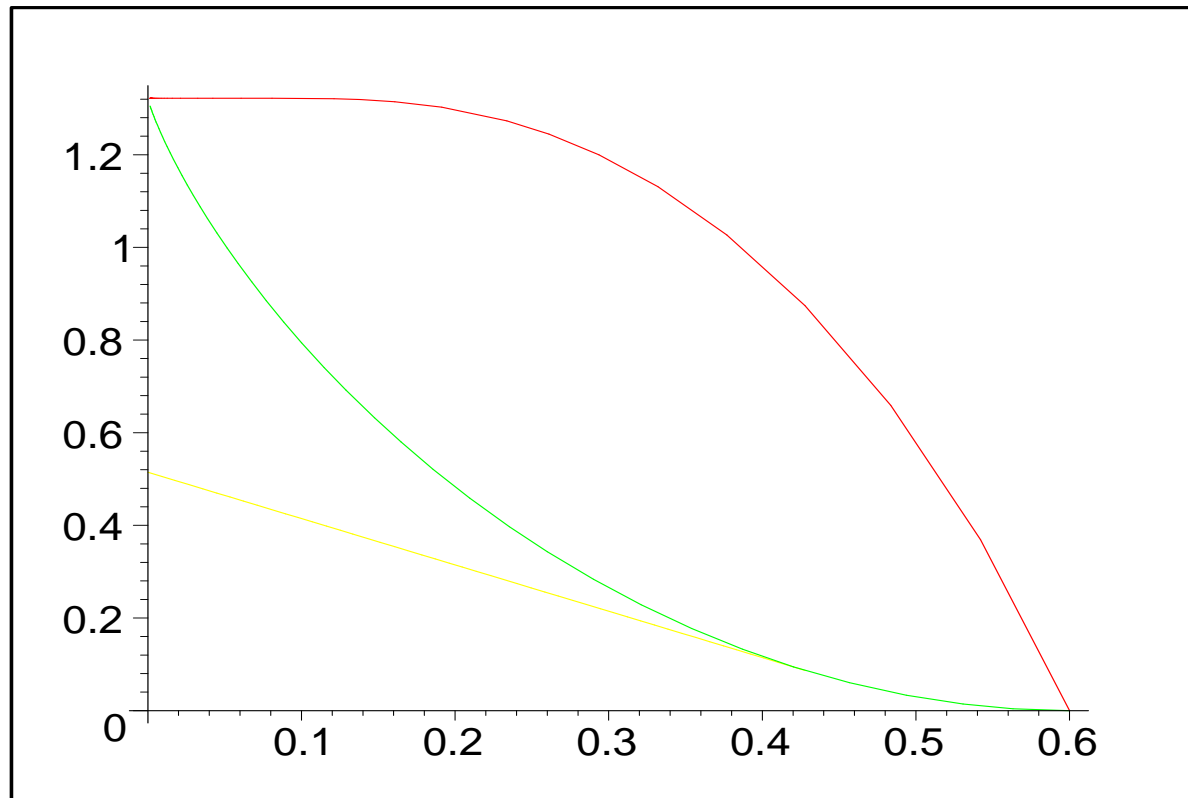
$$P(D \geq d) = P(L > \frac{d}{2}) = K \left( \frac{\delta}{1-\delta} \right)^d$$

- $\approx 0.584$  vs  $0.0294$  for block-coding!



# Compare BEC feedback-delay reliability $\alpha$ to classical $E_{sp}$

$$R(\alpha) = \frac{\alpha}{\alpha + \log_2\left(\frac{1-\delta}{1-\delta 2^\alpha}\right)}$$



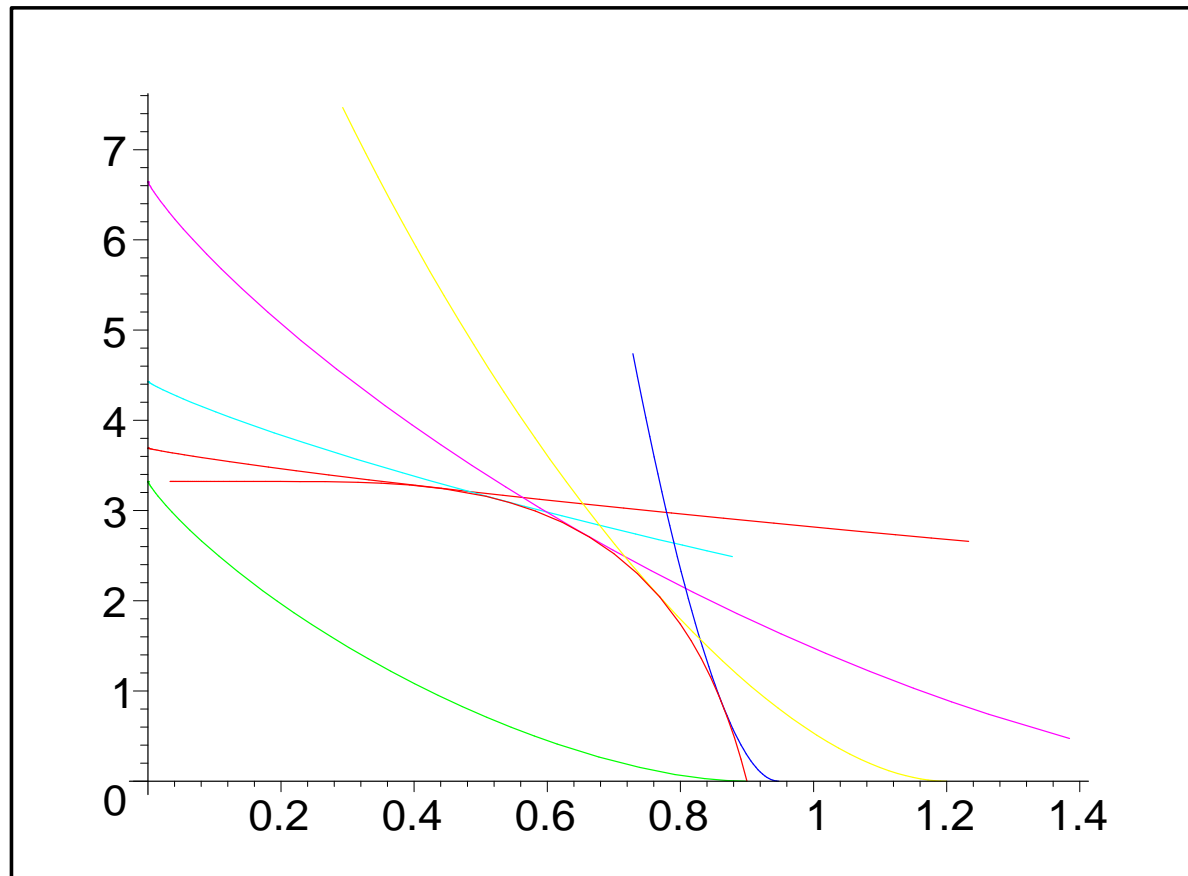
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## Using $E_{sp}$ to bound $\alpha^*$ in general

- Use a rate  $R$  anytime-code to make a block code of rate  $R' = (1 - \lambda)R$  where  $\lambda \in [0, 1]$ .
  - Take  $R'n$  bits of data and consider them the first bits to arrive at the anytime encoder.
  - For the rest of the data bits (taking time  $\lambda n$ ), just choose 0.
- The block error probability is bounded by  $K2^{-\alpha\lambda n}$  which can not exceed the sphere-packing bound  $2^{-E_{sp}((1-\lambda)R)n}$

$$\alpha^*(R) \leq \frac{E_{sp}((1-\lambda)R)}{\lambda}$$

## Upper bound tight for the BEC with feedback



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## Bound for symmetric DMCs

Minimize over  $\lambda$  for symmetric DMCs to sweep out frontier by varying  $\rho > 0$ :

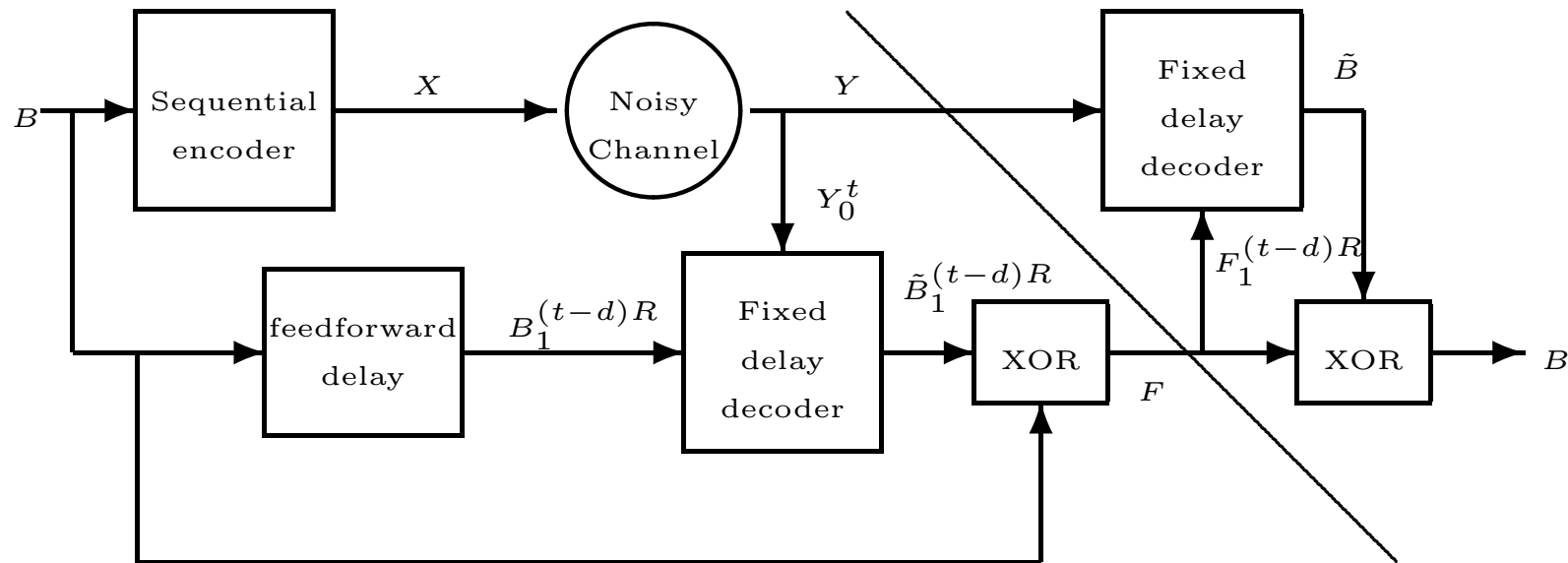
$$R(\rho) = \frac{E_a^+(\rho)}{\rho}$$

$$E_a^+(\rho) = -\max_q \log_2 \sum_j \left( \sum_i q_i p_{ij}^{\frac{1}{1+\rho}} \right)^{1+\rho}$$

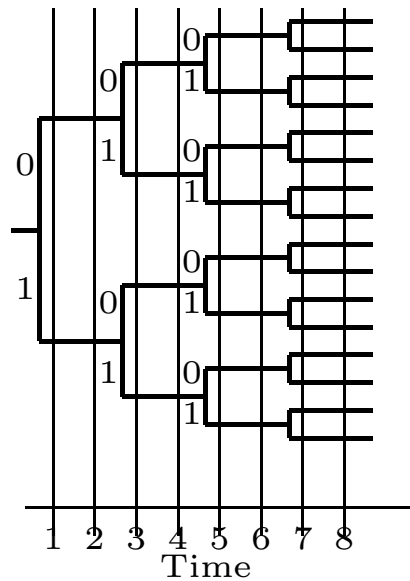
The same basic form as the sphere-packing bound, but concave  $\cap$  instead of convex  $\cup$ .

## Reliability: from general coding or from feedback?

- Without feedback:  $E_{sp}(R)$  continues to be a bound. (Pinsker)
- Consider a code with target delay  $d$ 
  - Use it to construct a block-code with blocksize  $n \gg d$
  - Genie-aided decoder: has the truth of all bits before  $i$
  - Error events for genie-aided system depend only on last  $d$
  - Apply a change of measure argument



## Anytime codes without feedback?



Infinite binary tree, with iid random labels:

- Choose a path through the tree based on data bits
- Transmit the path labels through the channel

- Can implement with time-varying convolutional code.
- Decoder does ML decoding
  - Disjoint paths are pairwise indep to true path.
  - $E_r(R)$  analysis applies.
- Achieves  $\alpha = E_r(R)$  for every  $d$  for all  $R < C$

## Open questions

- Value of delayed or noisy feedback?
- Bounds for  $(\vec{R}, \vec{\alpha})$  regions with/without feedback?
- Distributed problems — many sensors and/or many controllers.
  - Anytime MAC, Broadcast, and Slepian-Wolf exist.
  - Should yield new distributed sufficient conditions.
- Interaction with unmodeled dynamics
- **Performance**
  - Tatikonda's  $D_{seq}(R)$  bounds are still the best we have.
  - Anytime tells us more about the  $\infty$  vs finite boundary, but not about performance within the “finite” region.
  - For non-causal estimation of unstable processes, performance is due to residual capacity left over after anytime requirement is met.