

Control over Lossy Networks:

The Interplay Between Coding and Control

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Workshop on Interactive Information Theory

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Discovery*

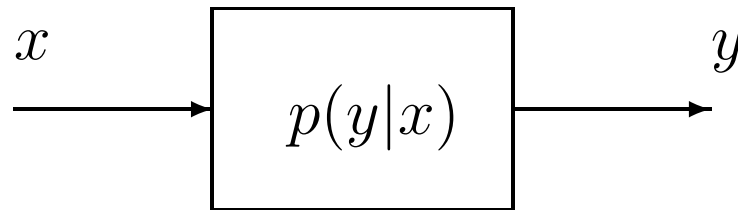
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Outline

- **Introduction**
 - information theory and coding: algebraic and graph-based codes
 - control theory: LQG control
- **Interplay between control and communication**
 - coding for interactive communications: tree codes
 - control over noisy channels: anytime capacity
 - estimation over lossy channels
- **Construction of linear tree codes**
 - existence with high probability
 - efficient decoding for erasure channels
 - examples
- **Conclusion**
 - future work and open problems

Single-User Information Theory

Single-user information (Shannon 1948) deals with the study of the fundamental limits of reliable information transmission between a sender and a receiver over a noisy channel



$$C = \max_{p_X(\cdot)} \{H(x) + H(y) - H(x, y)\}$$

Key idea: Block coding

- the behavior of the channel over a single use is unpredictable
- but the behavior over many channel uses is:
 - if the channel introduces errors with probability p , say, over $n \gg 1$ channel uses it will introduce $\approx np$ errors

Coding Theory

- start with $b = \{b_i\}_{i=1}^m$ bits (the message)
- map them to $c = \{c_i\}_{i=1}^n$ bits (encoding, rate = $\frac{m}{n}$)
- the set of all codewords, c , is denoted by \mathcal{C} ($|\mathcal{C}| = 2^m$)
- if $c = \{c_i\}_{i=1}^n$ is transmitted across the channel and $y = \{y_i\}_{i=1}^n$ is received, then the maximum likelihood decoder is

$$\hat{c} = \arg \max_{c \in \mathcal{C}} p(y|c)$$

Shannon showed that for all rates $\frac{m}{n} < C$, *there exists a sequence of codes*, such that

$$\lim_{n \rightarrow \infty} P(\hat{c} \neq c) = 0.$$

Information Theorists Live in Asymptopia

This is nice theory and an elegant result. However,

- it may require unlimited computational resources at the transmitter and receiver (encoding and decoding may need exponential time)
- it assumes asymptotically long delays ($n \rightarrow \infty$)
 - encoding can be done only *after* all the bits $\{b_i\}_{i=1}^n$ are available
 - decoding can be done only *after* all the outputs $\{y_i\}_{i=1}^n$ are observed

In Practice...

- one cares about the probability-of-error as a function of the rate and the length of the code (error exponents)
- one cares about codes that can be efficiently encoded and decoded
 - algebraic codes
 - * Reed-Solomon, Reed-Muller, algebraic geometry
 - * Berlekamp-Massey, list-decoding (Guruswamy-Sudan), etc.
 - graph-based codes
 - * turbo codes, LDPC codes, expander codes
 - * message-passing, bit-flipping, LP decoding, etc.
 - polar codes

In summary, after 60 years of work, we have practical codes that come close to the Shannon limits in many cases.

A Critique...

An early criticism (albeit philosophical) of information was that it *did not involve time*

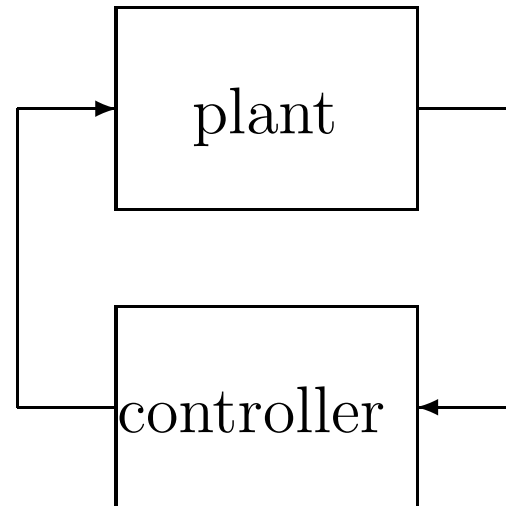
- the “information” obtained about knowledge of an event, depends only on the probability of that event

$$\log \frac{1}{p},$$

not on *when* this knowledge is revealed or when we want to take action on this knowledge

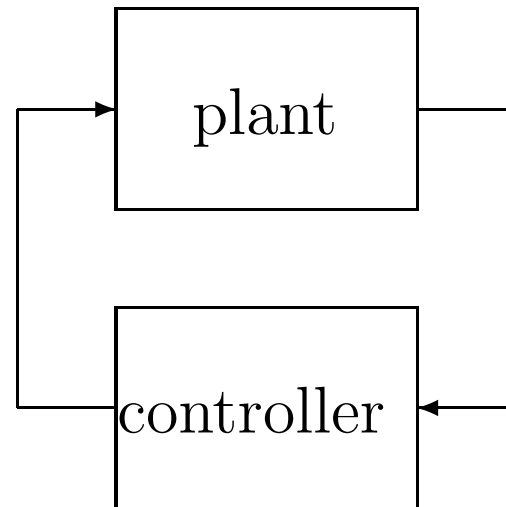
- issue never quite resolved (things like directed mutual information, or the entropy rate of a random process don't quite cut it)
- problem is that information-theoretic quantities often require some form of ergodicity to have operational significance

Control Theory



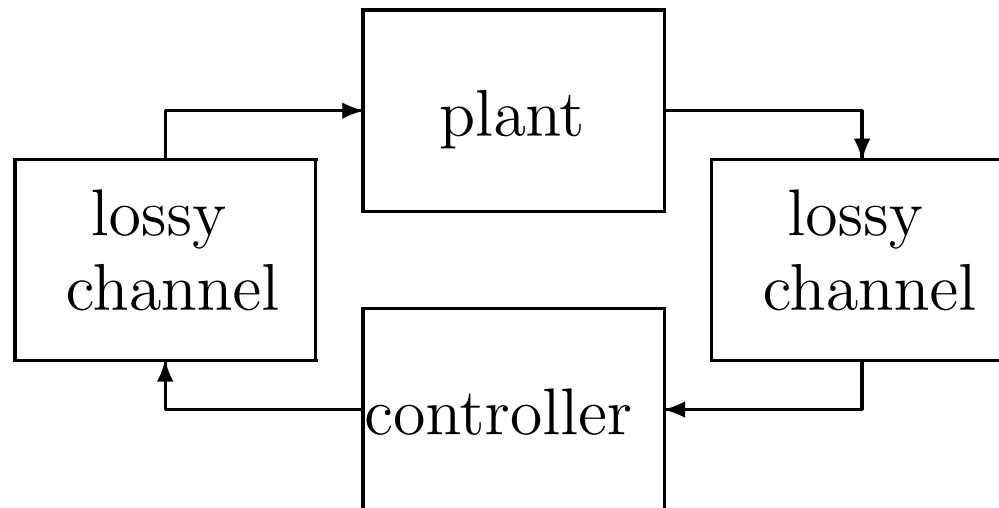
- in control theory we observe the output of a dynamical system (plant) and design a controller to regulate its behavior
- controller needs to react and generate control signals on the fly (*in real-time*)
- the introduction of delay can result in loss of performance and/or instability

Control Theory



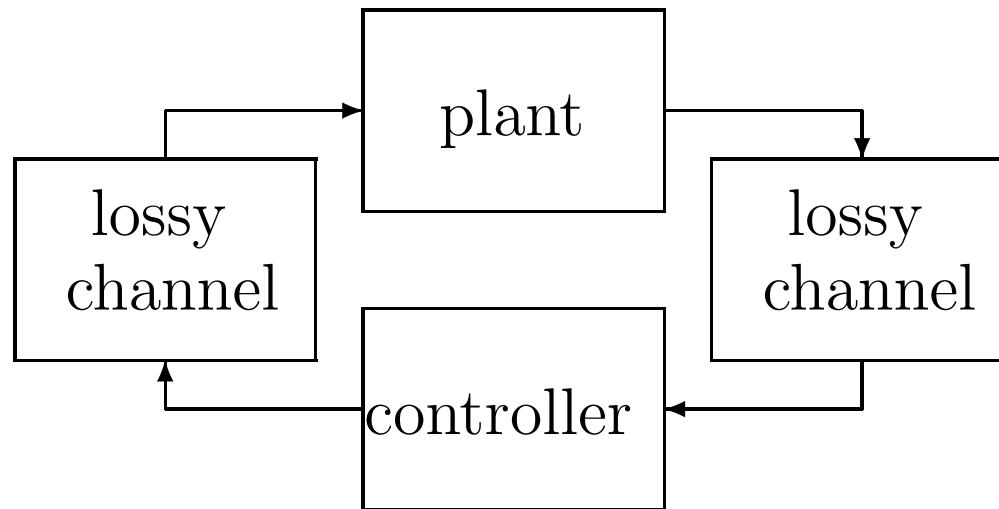
- very rich theory has been developed (especially in the LTI case)
 - LQG control, H^∞ control, Kalman filtering, separation principle
- virtually no interaction with information theory
 - plant and controller often co-located, no measurement loss

But what if...?



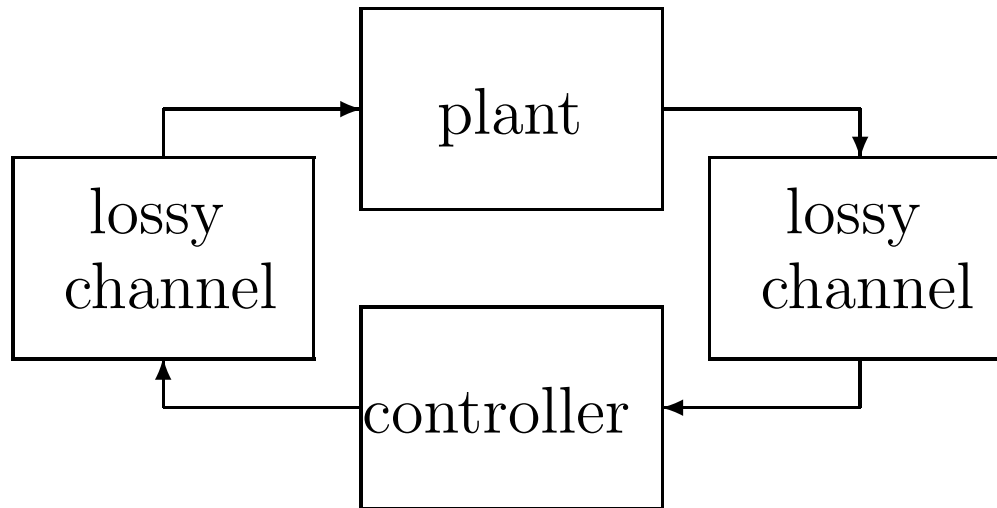
- increasingly we have applications where systems (autonomous agents, sensor/actuator networks, smart grid, etc.) are remotely controlled and where measurement and control signals are transmitted across noisy channels
- conventional channel coding does not work - the ensuing delay may lead to instability

Living with the Noisy Channels...?



- if the noisy channels are erasure channels, Sinopoli et al (2005) showed that if the erasure probability is high enough then the closed loop system will be unstable
- similar results hold for other classes of channels

What to Do?



- the problem is that if we cannot tolerate large delays, we cannot make the noisy channels reliable
- but do we need to do that?
- what do we need to guarantee the stability of the closed loop system?

Coding for Interactive Communications

Consider a two-party communication system

Alice, x

Bob, y

$$s_1 = f_1(x)$$

----->

$$s_2 = f_2(y, s_1)$$

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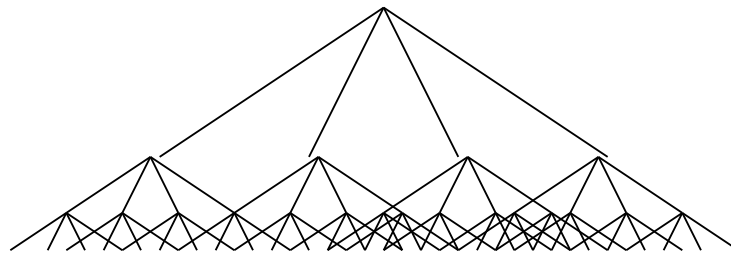
$$s_3 = f_3(x, s_1, s_2)$$

----->

⋮

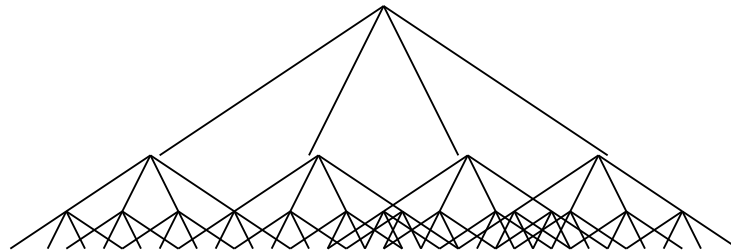
Can one do this reliably over noisy links?

Tree Codes (Schulman, 1993)



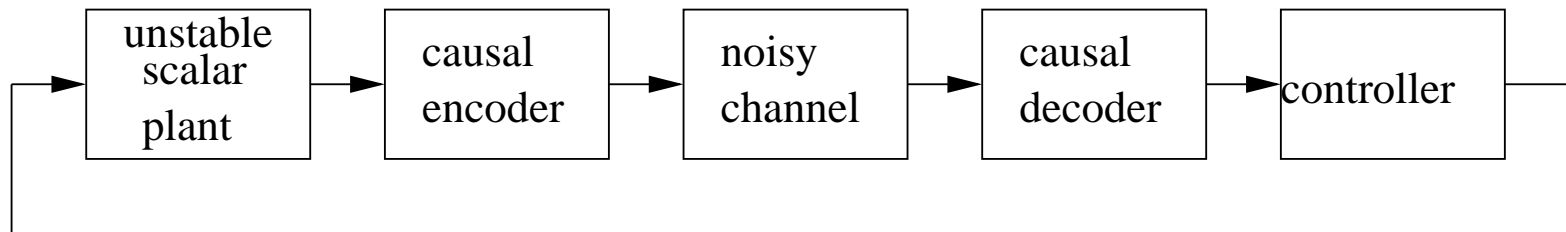
- semi-infinite d -ary tree
- each edge labeled by a symbol in an alphabet of size $d' > d$
- maps a sequence $\{s_i\}_{i=1}^{\infty}$ to a sequence $\{c_i\}_{i=1}^{\infty}$, where $s_i \in \{0, 1, \dots, d-1\}$ and $c_i \in \{0, 1, \dots, d'-1\}$
- represents a *causal* code; each path is a codeword; $R = \frac{\log d}{\log d'}$

Tree Codes (Schulman, 1993)



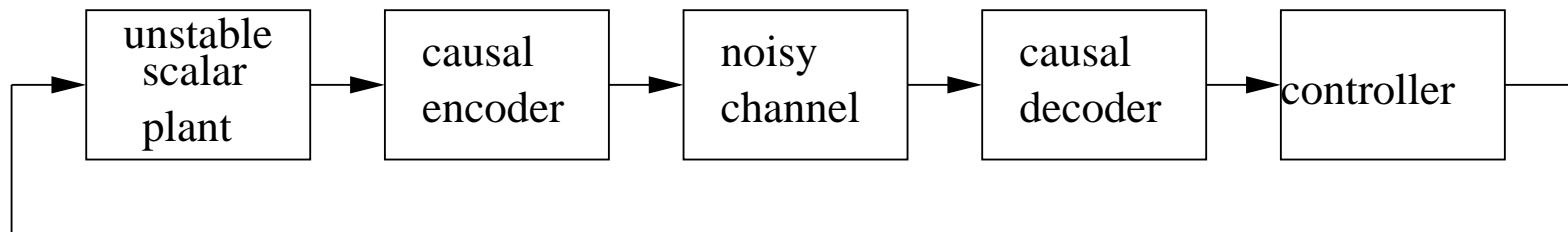
- for *every* pair of paths with a common ancestor and length n , say, we require that the "Hamming distance" between the paths be *at least a fixed proportion of n*
- Schulman proved the existence of tree codes
- along with ML decoding, allows reliable interactive communication over a noisy link
- **Problem:** No explicit constructions; no tractable decoding; existence result is not with high probability

Anytime Capacity (Sahai, 2001)



- scalar unstable LTI system
- noisy channel from plant output to controller
- each measured output quantized to k bits; *causally* encoded and transmitted across channel
- controller attempts to *causally* decode transmitted bits, estimate state of the system and generate a control signal to stabilize the plant

Anytime Capacity (Sahai, 2001)



- how big should k be?
- given that we cannot reliably recover the transmitted bits, can we even stabilize the system?
- what sort of fidelity do we need?

Toy Example: Tracking an Unstable Plant

$$x_{i+1} = ax_i + w_i, \quad |a| > 1, \quad w_i \in \{-1, 1\} \text{ (unknown)}$$

Assume the initial state $x_0 = 0$ is known to the encoder and controller.

- clearly, at each time instant, the encoder should try to convey 1 bit of information to the controller indicating whether $w_i = 1$ or $w_i = -1$
- the encoder will causally encode this sequence of bits $\{b_i\}_{i=0}^{\infty}$ and send them across the channel
- the decoder, at each time instant i , will attempt to decode the *entire* bit sequence $\{b_j\}_{j=0}^i$ and obtain $\{\hat{b}_{j|i}\}_{j=0}^i$

Toy Example: Tracking an Unstable Plant

Define

$$P_e(i, d) = \text{Prob}(\hat{b}_{j|i} = b_j, \forall j < i - d, \hat{b}_{i-d|i} \neq b_{i-d})$$

This is the probability that the first error happens d time steps in the past.

- Then the mean-square error is bounded by

$$E(x_{i+1} - \hat{x}_{i+1|i})^2 \leq \sum_{d=1}^i \left(\frac{a^d - 1}{a - 1} \right)^2 P_e(i, d) < \frac{1}{(a - 1)^2} \sum_{d=1}^{\infty} |a|^{2d} P_e(i, d).$$

- Clearly, if there exists K , ϵ and Δ , such that for all i and $d > \Delta$:

$$P_e(i, d) < K|a|^{-2d-\epsilon},$$

we will have $E(x_{i+1} - \hat{x}_{i+1|i})^2 < \infty$, for all i , i.e., we will have *mean-square stability*.

- **Remark:** For mean absolute stability, we just need

$$P_e(i, d) < K|a|^{-d-\epsilon}$$

Conclusion: We do not need arbitrary reliability. Only a reliability that decays appropriately exponentially fast with the delay.

Anytime Capacity

- *Definition:* A channel will be said to have "anytime capacity" $C_{any}(\lambda)$, for some parameter $\lambda > 1$, if for all rates $R < C_{any}(\lambda)$, there exists causal encoding and decoding schemes such that

$$P_e(i, d) < K\lambda^{-d-\epsilon}, \quad \forall i, \forall d > \Delta$$

- *Theorem:* Consider a scalar LTI system

$$\begin{cases} x_{i+1} &= ax_i + w_i + u_i \\ y_i &= x_i + v_i \end{cases} \quad |\lambda| > 1$$

where w_i and v_i are *bounded* disturbances. Then to stabilize this system over a noisy channel it is necessary and sufficient that

1. $k > \log |\lambda|$
 2. $R = \frac{k}{n} < C_{any}(|\lambda|)$
- The theorem is based on the use of tree codes

Some Problems

This is an elegant result, but...

- there are no explicit constructions of tree codes with efficient decoding
- there is very little hope of actually computing $C_{any}(\lambda)$
 - this requires computing optimal error exponents for tree codes
 - even for block codes optimal error exponents have not been computed
- there are no "necessary and sufficient" conditions for systems with vector states

What to do...?

Linear Tree Codes

- linear codes can be represented by *generator* or, equivalently, *parity check* matrices
- a linear tree code will thus clearly have a lower triangular generator matrix

$$\begin{bmatrix} c_1 \\ c_2 \\ c_3 \\ \vdots \end{bmatrix} = \begin{bmatrix} G_{11} & & & & \\ G_{21} & G_{22} & & & \\ G_{31} & G_{32} & G_{33} & & \\ \vdots & \vdots & \vdots & \ddots & \end{bmatrix} \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ \vdots \end{bmatrix},$$

where $b_i \in GF_2^k$, $c_i \in GF_2^n$ and $G_{ij} \in GF_2^{n \times k}$

- equivalently, the parity check matrix can be made lower triangular

$$\begin{bmatrix} P_{11} & & & \\ P_{21} & P_{22} & & \\ P_{31} & P_{32} & P_{33} & \\ \vdots & \vdots & \vdots & \ddots \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \\ c_3 \\ \vdots \end{bmatrix} = 0$$

where $P_{ij} \in GF_2^{(n-k) \times n}$

- *Do linear tree codes exist?* Requires $P_e(i, d) < K\lambda^{-d-\epsilon}$, for all i and $d > \Delta$. This requires two union bounds (which kills things)—hence Schulman's approach

Toeplitz Linear Tree Codes

- trick is to make the code Toeplitz

$$\begin{bmatrix} G_0 & & & \\ G_1 & G_0 & & \\ G_2 & G_1 & G_0 & \\ \vdots & \vdots & \vdots & \ddots \end{bmatrix}$$

and

$$\begin{bmatrix} P_0 & & & \\ P_1 & P_0 & & \\ P_2 & P_1 & P_0 & \\ \vdots & \vdots & \vdots & \ddots \end{bmatrix}$$

- this makes the code "look the same" at all times i , and so we avoid the union bound over i

Existence of Tree Codes with High Probability

Theorem: Choose the entries of the entries of the matrices $\{G_i\}_{i=1}^{\infty}$ independently from Bernouli($\frac{1}{2}$) and consider a binary-input channel with Bhattacharya parameter

$$\zeta = \int_{-\infty}^{\infty} \sqrt{p(y|b=1)p(y|b=0)} dy.$$

Then with probability $1 - 2^{-\Omega(n\Delta)}$, for all rates satisfying

$$R < 1 - \log(1 + \zeta),$$

there exists a K such that the probability of ML decoding satisfies

$$P_e(i, d) < K2^{-\beta d}, \quad \forall i, \forall d > \Delta$$

where

$$\beta < H^{-1}(1 - R) \left(\log \frac{1}{\zeta} + \log(2^{1-R} - 1) \right).$$

- The theorem holds exactly if we choose the entries of the $\{P_i\}_{i=0}^{\infty}$ from Bernouli($\frac{1}{2}$)
- For the BSC(p), the Bhattacharya parameter is $2\sqrt{p(1-p)}$ and therefore

$$R < 1 - \log(1 + 2\sqrt{p(1-p)}) = 1 - 2\log(\sqrt{p} + \sqrt{1-p}).$$

- For the BEC(ϵ), the Bhattacharya parameter is ϵ and therefore

$$R < 1 - \log(1 + \epsilon).$$

But What to do About ML Decoding?

- to get the error performance we need, we must do ML decoding *at each time instant*
- for block codes doing this even once is too hard....

But not for Erasure Channels

- Consider a random block linear code with $(N - K) \times N$ parity check matrix P ($R = \frac{K}{N}$)
- Suppose the codeword c is transmitted and partition it onto the observed entries c_o and the erased entries c_e , i.e., $c = \begin{bmatrix} c_o \\ c_e \end{bmatrix}$.
- Now due to the parity check condition

$$Pc = \begin{bmatrix} P_o & P_e \end{bmatrix} \begin{bmatrix} c_o \\ c_e \end{bmatrix} = 0, \text{ we have}$$

$$P_e c_e = P_o c_o$$

-

$$P_e c_e = P_o c_o$$

If the erasure probability of the channel is ϵ , then c_e will have size $\approx N\epsilon$

- If $R = \frac{K}{N} < C = 1 - \epsilon$, then $N - K > N\epsilon$ and the system of linear equations $P_e c_e = P_o c_o$, will, with high probability, have a unique solution.

Conclusion: For erasure channels ML decoding is simply matrix inversion

But What About the Lower Triangular Case?

- for tree codes the matrices P_e and P_o are lower triangular
- therefore even though the matrix P_e has more rows/equations ($N - K$) than columns/unknowns ($\approx N\epsilon$), the system of equations

$$P_e c_e = P_o c_o,$$

will most likely *not* have a unique solution (otherwise tree codes would have the same performance of block codes!)

- however, if the tree code corrects all errors above a delay of d , this means that if we partition $c_e = \begin{bmatrix} c_{e1} \\ c_{e2} \end{bmatrix}$, where c_e are all the erased entries with delay more than d , we must have

$$P_e c'_e = P_e c''_e \quad \text{implies} \quad c'_{e1} = c''_{e2}$$

- partitioning P_e , P_o and c_o similarly, we have

$$\begin{bmatrix} P_{e11} & \\ P_{e21} & P_{e22} \end{bmatrix} \begin{bmatrix} c_{e1} \\ c_{e2} \end{bmatrix} = \begin{bmatrix} P_{o11} & \\ P_{o21} & P_{o22} \end{bmatrix} \begin{bmatrix} c_{o1} \\ c_{o2} \end{bmatrix}$$

or

$$\begin{bmatrix} P_{e11}c_{e1} \\ P_{e21}c_{e1} + P_{e22}c_{e2} \end{bmatrix} = \begin{bmatrix} P_{o11}c_{o1} \\ P_{o21}c_{o1} + P_{o22}c_{o2} \end{bmatrix}$$

pre-multiplying the second set of equations by P_{e22}^\perp , the orthogonal complement of P_{e22} yields

$$\begin{bmatrix} P_{e1} \\ P_{e22}^\perp P_{e21} \end{bmatrix} c_{e1} = \begin{bmatrix} P_{o11}c_{o1} \\ P_{e22}^\perp P_{o21}c_{o1} + P_{e22}^\perp P_{o22}c_{o2} \end{bmatrix}$$

But this latter system of equations must have a *unique solution* for c_{e1} .

An Efficient Algorithm

1. suppose at time i the bits up to delay d have not yet been decoded
(this happens with probability $P_e(i, d) < K\lambda^{-d}$)

2. for these bits, partition $c = \begin{bmatrix} c_e \\ c_o \end{bmatrix}$ and $P = \begin{bmatrix} P_e & P_o \end{bmatrix}$

3. starting with delays $d' = 1, 2, \dots, d$ check whether the matrix

$$\begin{bmatrix} P_{e1} \\ P_{e22}^\perp P_{e21} \end{bmatrix}$$

has full column rank

4. if so, solve for c_{e1} in the system of equations

$$\begin{bmatrix} P_{e1} \\ P_{e22}^\perp P_{e21} \end{bmatrix} c_{e1} = \begin{bmatrix} P_{o11} c_{o1} \\ P_{e22}^\perp P_{o21} c_{o1} + P_{e22}^\perp P_{o22} c_{o2} \end{bmatrix}$$

5. if this does not happen for any $d' = 1, 2, \dots, d$, go to next time instant

The expected complexity per time instant is constant:

$$\sum_{d=1}^{\infty} K' d^3 \lambda^{-d}$$

Furthermore, the probability that the complexity at any given time instant exceeds $O(d^3)$ decays as $O(\lambda^{-d})$.

Remark: With feedback, encoding can also be done with constant expected complexity.

A Scalar Example

- Take the scalar LTI system

$$\begin{cases} x_{i+1} & = & 2x_i + w_i + u_i \\ y_i & = & x_i + v_i \end{cases}$$

where w_i is uniform over $[-30, 30]$ and v_i is uniform over $[-1, 1]$.

- Suppose we want to stabilize this over an erasure channel with erasure probability $\epsilon = 0.3$ and that we have $n = 15$ bits per measurement at our disposal.
- We need an error exponent $2^{-\beta} < \frac{1}{2}$. Using the theorem we can see that we need a rate less than $R < 0.40$, i.e., we should quantize the measurements to at most $k = 5$ bits.

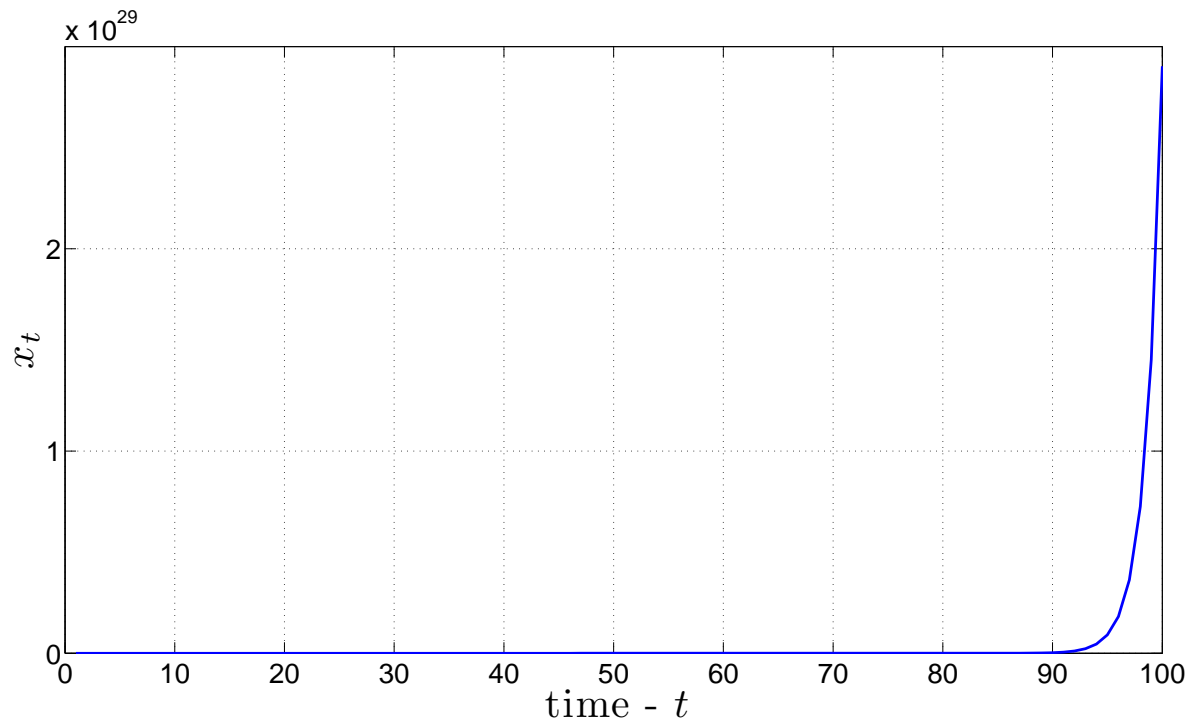


Figure 1: Open loop response.

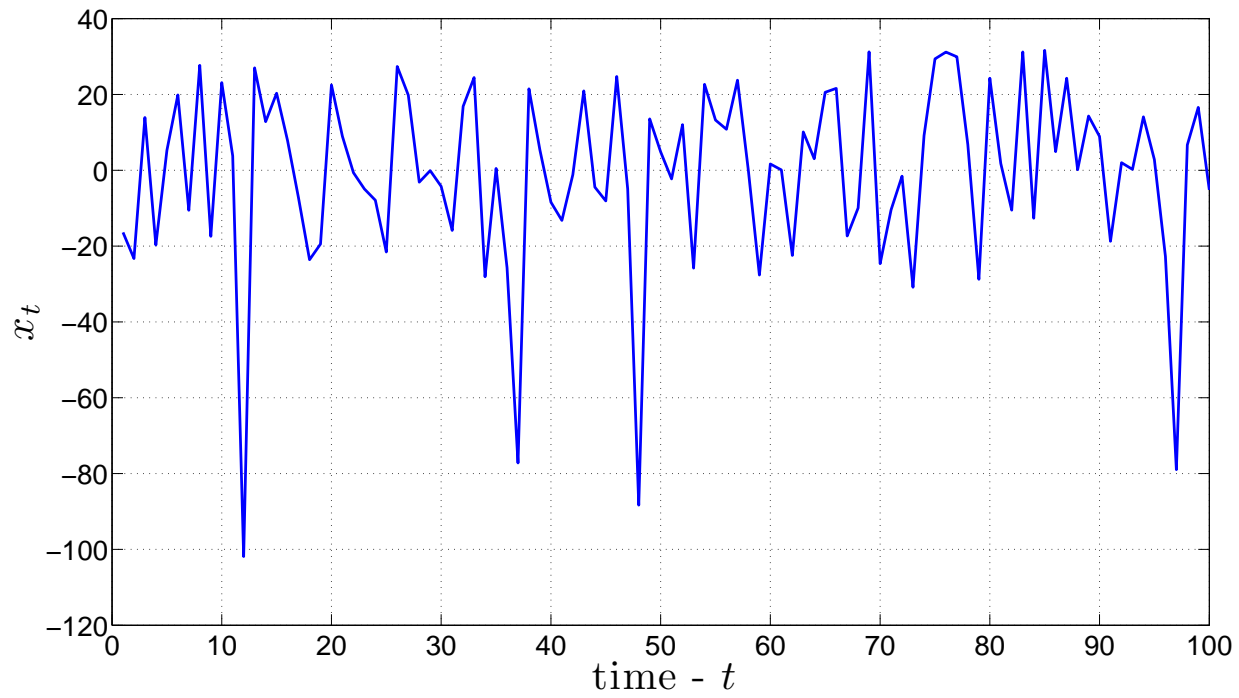


Figure 2: Closed loop response, $k = 5$.

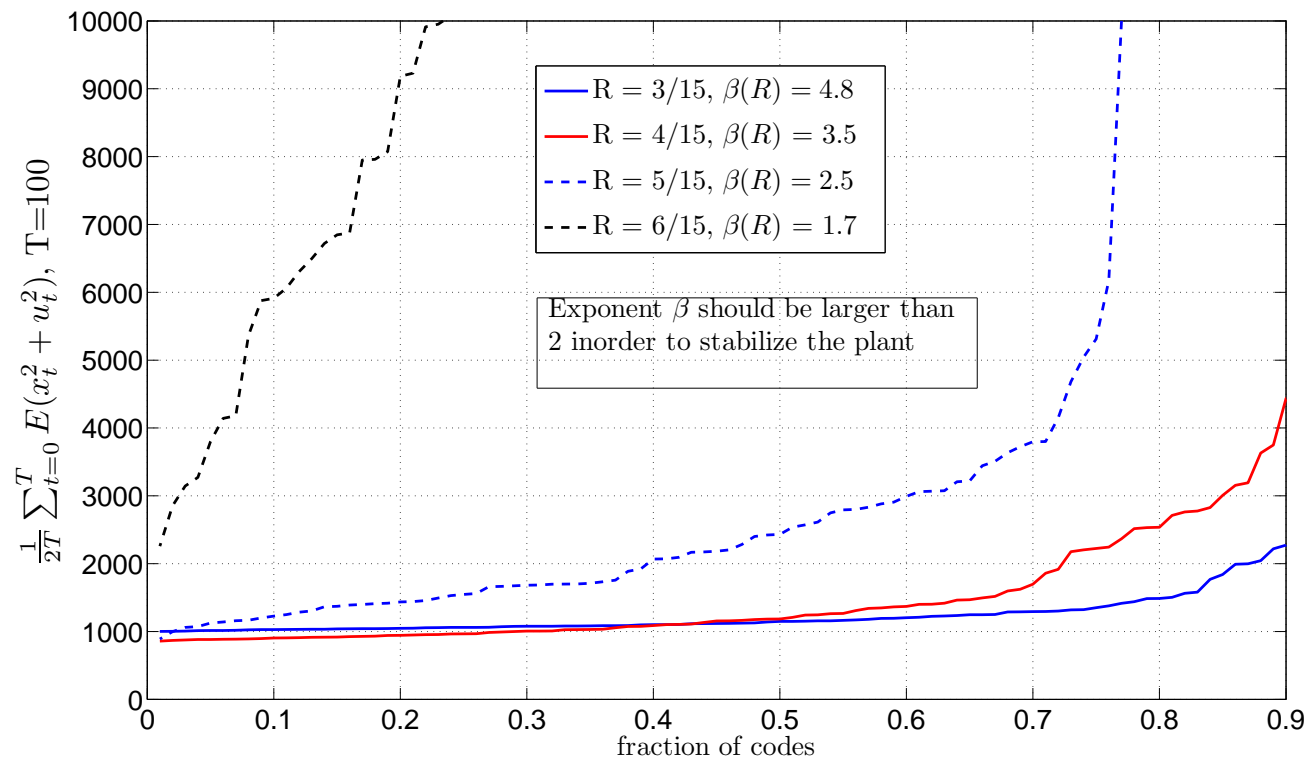


Figure 3: CDF of LQR costs for different realizations of the codes.

The Vector Case

Necessary and sufficient conditions to guarantee stability for LTI systems with vector state-space is not known. Here is a useful sufficient condition.

- Consider an LTI system with vector state-space, scalar measurement, and *bounded* system and measurement noise.
- Wlog write the state-space system in observer canonical form

$$\left\{ \begin{array}{l} x_{i+1} \\ y_i \end{array} \right. = \begin{bmatrix} -a_1 & 1 & 0 & \dots & 0 \\ -a_2 & 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \ddots & \vdots \\ -a_n & 0 & 0 & \dots & 1 \\ 1 & 0 & 0 & \dots & 0 \end{bmatrix} x_i + w_i + u_i$$

Note that the a_i are the coefficients of the characteristic polynomial of the system matrix.

- Let $\lambda > 1$ be the smallest positive number such that the matrix

$$\begin{bmatrix} \frac{|a_1|}{\lambda} & 1 & 0 & \dots & 0 \\ \frac{|a_2|}{\lambda} & 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \ddots & \vdots \\ \frac{|a_n|}{\lambda} & 0 & 0 & \dots & 1 \end{bmatrix}$$

is stable.

- Then we can stabilize the system (in a mean-square sense) by appropriately quantizing each measurement with k bits and using a tree code, such that
 1. $k > \log \lambda$
 2. $P_e(i, d) < K|\lambda_{max}|^{-2d-\epsilon}$ for all i and $d > \Delta$, where $|\lambda_{max}|$ is the largest eigenvalue of the system matrix (in absolute value).

A Vector Example

$$\begin{aligned}x_{i+1} &= \begin{bmatrix} 2 & 1 & 0 \\ 0.25 & 0 & 1 \\ -0.5 & 0 & 0 \end{bmatrix} x_i + w_i + u_i \\ y_i &= \begin{bmatrix} 1 & 0 & 0 \end{bmatrix} x_i + v_i\end{aligned}$$

where w_i and v_i are truncated $N(0, 1)$ normals to lie in $[-2.5, 2.5]$.

$\lambda_{max} = 2$.

- We would like to stabilize the plant over an erasure channel with $\epsilon = 0.3$.
- We have $n = 15$ bits per measurement available.
- We need an exponent $< \frac{1}{2}$: an application of the theorem shows that we need a rate $R < 0.43$, hence $k < 7$.

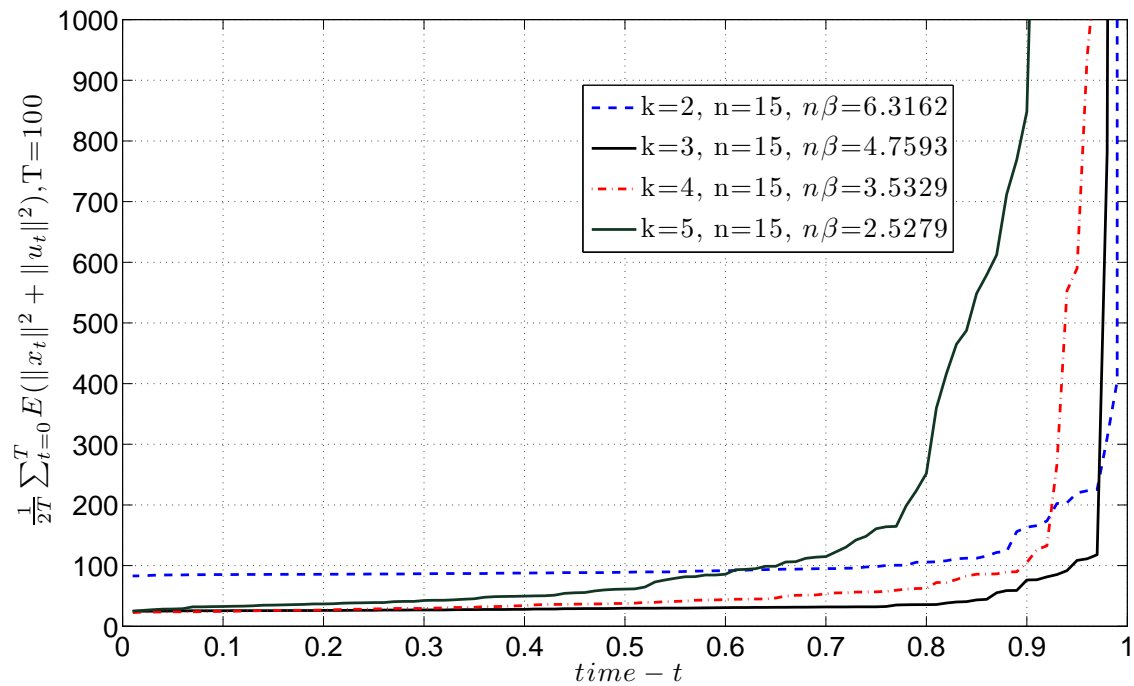


Figure 4: CDF of LQR costs for different realizations of the codes.

Some Remarks

- We have developed a universal and efficient method for stabilizing plants driven by *bounded* noise over erasure channels
- This is perhaps the most practically interesting case: most systems will probably quantize their measurements to some number of bits, put them in packets and send them across a lossy network (where packets will either be received or dropped)
- Not clear how to deal with unbounded noise (say, Gaussian). Seems to require perfect feedback. Not clear if this is important in practice.
- Stabilizing a plant is the first step. Optimizing performance is the next.
 - this will require studying the trade-offs between control and communication resources
 - should we quantize coarsely, but heavily protect the bits, or quantize finely and moderately protect the bits?

Other Channels

- It would be interesting to develop efficiently-decodable tree codes for other types of channels, especially, the BSC and the AWGNC.
- These appear to be much more challenging, since ML decoding is out of the question.
 - even in the block coding case, the codes that achieve capacity over these channels (such as LDPCs or polar codes) do not do so with an error exponent
 - for example, polar codes approach capacity with a probability-of-error $e^{-\alpha\sqrt{N}}$
- We care about having an error exponent much more than the rate
 - in the block coding case, decoders that achieve an error exponent are those that can correct a fixed fraction of errors
 - Reed-Solomon codes; LDPC and expander codes with bit-flipping and/or LP decoding

Conclusion

- Traditional information theory lives in Asymptopia—not appropriate for real-time constraints
- Control has long ignored information theory...
- Controlling an unstable plant over a noisy channel is one place where the two must meet
 - the key object is a “tree code” (essentially a causal code), rather than a block code
 - the key criterion is the interplay between the rate and the decay of the probability of error as a function of the delay (anytime capacity)

- For the first time:
 - showed the existence (with high probability) of linear tree codes for a wide class of channels
 - developed an efficiently decodable tree code for erasure channels and demonstrated the efficacy of the method
- This opens up many possibilities for controlling distributed systems over noisy communication links and lossy networks
- It is important to study how best to trade off control and communication resources to optimize system performance
- Developing efficiently-decodable tree codes for other classes of channels is an important and interesting open problem