

(Almost) Practical Tree Codes

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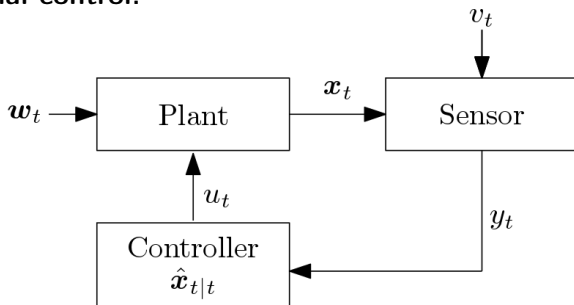
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Networked Control vs. Traditional Control

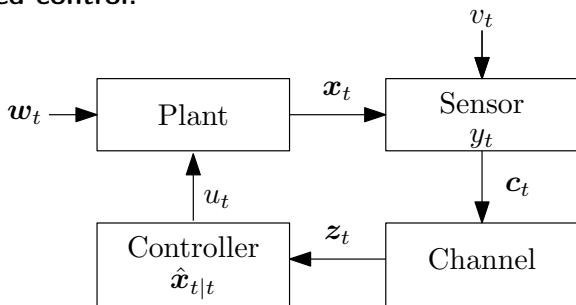
Traditional control:



- Observer and controller are co-located.
- Classical systems are hardwired and well crafted

Networked Control vs. Traditional Control

Networked control:

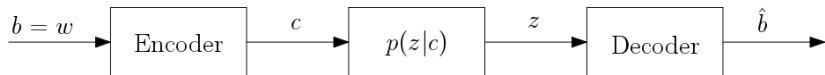


- Observer and controller are not co-located:
connected through noisy link
- Suitable for new remote applications
(e.g., remote surgery, self-driving cars)

Motivating Example: Tracking a Random Walk [Sahai PhD'01]

$$x_{t+1} = \alpha x_t + w_t$$

- $\alpha > 1 \implies$ not stable!
- $w_t \in \{\pm 1\}$
- We wish to track x_t with bounded expected distortion
- If tracking is possible, stability usually follows
- Allows to distill the coding problem (no quantization)



Distortion requirement

$$\mathbb{E} \left[(x_t - \hat{x}_t)^2 \right] < \infty, \quad \forall t$$

Motivating Example: Tracking a Random Walk [Sahai PhD'01]

- $\hat{b}_{t-d|t}$ – Estimate of b_{t-d} at time t
- Probability of first error event at time $t - d$:

$$P_e(t, d) \triangleq \Pr \left(b_{t-d} \neq \hat{b}_{t-d|t}, \forall \delta > d, b_{t-\delta} = \hat{b}_{t-\delta|t} \right)$$

$$\mathbb{E} \left[(x_t - \hat{x}_{t|t})^2 \right] \propto \sum_{d=1}^t P_e(t, d) \alpha^{2d} = \sum_{d=1}^t P_e(t, d) 2^{2 \log \alpha \cdot d} < \infty$$

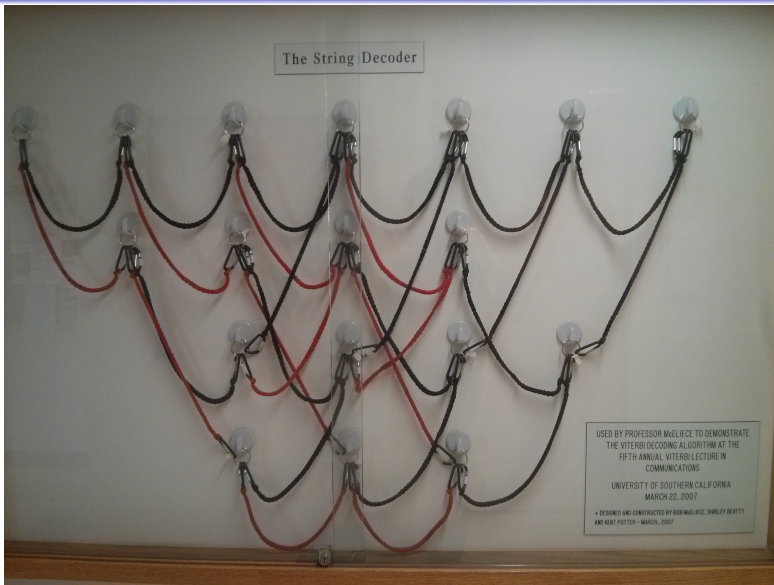
Error probability profile: Anytime-reliable code

$$P_e(t, d) < 2^{-(2 \log \alpha + \epsilon)d}, \quad \forall t, d_0 < d < t$$

Larger moments

Higher exponent \implies Cannot stabilize all moments!

Tree / Anytime-Reliable Codes



Anytime-Reliable Codes: Basics

Error probability profile

$$P_e(t, d) < 2^{-(2 \log \alpha + \epsilon)d}, \quad \forall t, d_0 < d < t$$

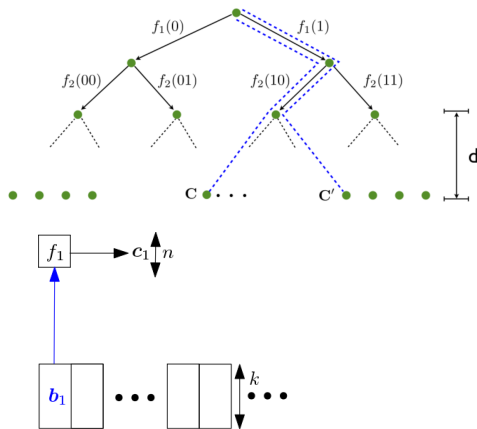
How to generate such a code?

$$\mathbf{c}_1 = f_1(\mathbf{b}_1)$$

$$\mathbf{c}_2 = f_2(\mathbf{b}_1, \mathbf{b}_2)$$

$$\vdots$$

$$\mathbf{c}_t = f_t(\mathbf{b}_1, \mathbf{b}_2, \dots, \mathbf{b}_t)$$

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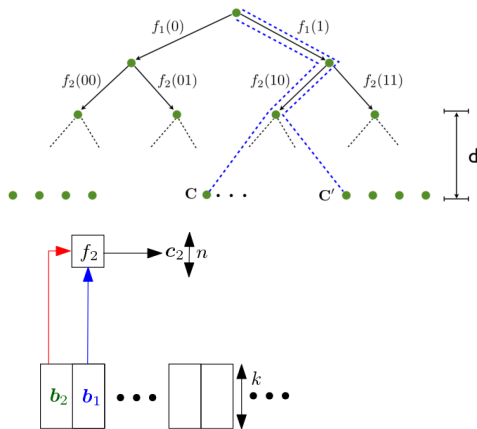
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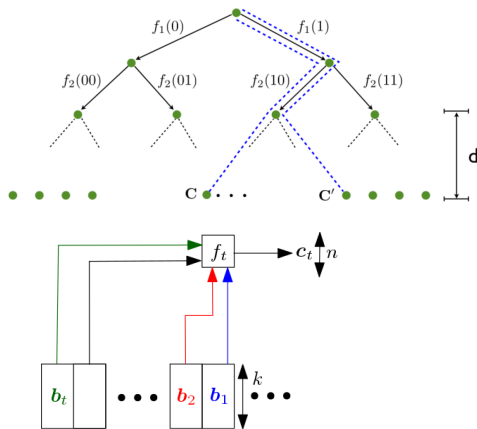
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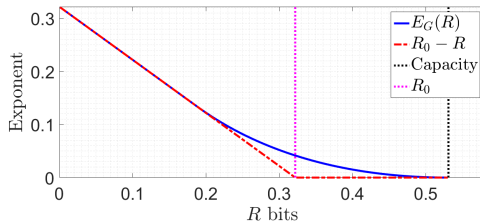
Anytime-Reliable Codes as Convolutional Codes

Random **time-varying** convolutional-code ensemble [Viterbi, Yudkin, Zigangirov, Shulman–Feder, ...]

- Most results assume infinite stream (\gg delay-line length)
- We wish to recover a bit using subsequent nd output symbols
- The random time-varying CC ensemble achieves:

$$\mathbb{E}[P_e(t, d)] \leq 2^{-E_G(R)nd}$$

- $E_G(R) > 0$ for $R < C$ – Gallager's error exponent



Anytime-Reliable Codes as Convolutional Codes

Good ensemble performance \Rightarrow Good specific code performance?

- Block codes: Yes, with high probability!
- Anytime reliable-code?
- Such a code exists [Schulman IT'96], but **not w.h.p.** 😞
(Proof requires min-distance \propto delay)
- **How to construct a good anytime-reliable code?**

Ensemble Performance \Rightarrow Specific Code Performance?

Ensemble performance

$$\mathbb{E}[P_e(t, d)] \leq 2^{-E_G(R)nd}$$

Specific d and t

Using Markov's inequality:

$$\Pr\left(P_e(t, d) \geq 2^{-[E_G(R)-\epsilon]nd}\right) \leq \frac{\mathbb{E}[P_e(t, d)]}{2^{-[E_G(R)-\epsilon]nd}} = 2^{-\epsilon nd}$$

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Linear Time-Invariant Codes [Sukhavasi–Hassibi ISIT'11]

Linear time-variant code

$$\mathbf{G} = \begin{bmatrix} \mathbf{G}_{1,1} & \mathbf{0} & \cdots & \cdots & \cdots \\ \mathbf{G}_{2,1} & \mathbf{G}_{2,2} & \mathbf{0} & \cdots & \cdots \\ \vdots & \vdots & \ddots & \ddots & \cdots \\ \mathbf{G}_{t,1} & \mathbf{G}_{t,2} & \cdots & \mathbf{G}_{t,t} & \mathbf{0} \\ \vdots & \vdots & \vdots & \vdots & \ddots \end{bmatrix}$$

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Using the union bound:

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$$= \frac{2^{-\epsilon nd_0}}{1 - 2^{-\epsilon n}}$$

Linear Time-Invariant Codes

- Random LTI convolutional codes are anytime-reliable w.h.p. ✓
- But the exponent result was valid for time-variant codes
- Valid also for LTI codes [Schulman–Feder IT'00] ✓
 - (Proved independently in [Sukhavasi–Hassibi ISIT'11])
- **No gain** for general codes over LTI codes in this regime!
 - (Common setting of infinite decoding window: huge gain!)

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What about decoding?

Decoding of Linear Time-Invariant Codes

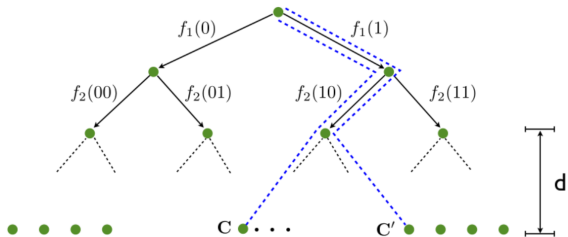
- All results assumed maximum-likelihood (ML) decoding
- ML complexity rises exponentially with t

Binary Erasure Channel (BEC)

- For LTI codes: ML = Solving linear equations
- What about other channels?

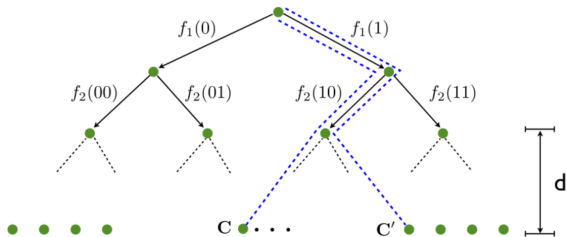
Sequential Decoding

- Before Viterbi algo.: Sequential decoding *de facto* standard
- Sequential decoding = class of algorithms
- Introduced originally in [Wozencraft '57] for **tree codes**
- Common to all: Explore only subset of (likely) codewords
- Most prominent variants: Stack and Fano's algorithms
- Proposed for general tree ensembles in [Sahai-Palaiyanur '05]



Sequential Decoding

- Fano's metric: $M(\mathbf{c}_1, \dots, \mathbf{c}_t) = \sum_{i=1}^{nt} \left[\log \frac{p(z_t | \mathbf{c}_t)}{p(z_t)} - \overbrace{B}^{\text{bias}} \right]$
- For ML decoding: $\arg \max_{\{\mathbf{c}_i\}} p(z_t | \mathbf{c}_t) = \arg \max_{\{\mathbf{c}_i\}} \left[\log \frac{p(z_t | \mathbf{c}_t)}{p(z_t)} - B \right]$
- For partial tree exploration: Fano's metric penalizes longer incorrect paths via bias B

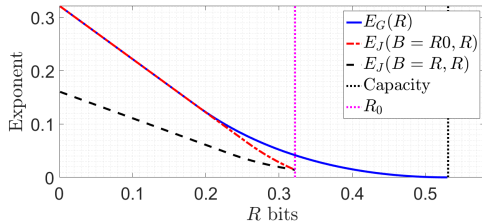


Sequential Decoding: Error Probability

Error probability of general conv. ensemble [Jelinek's Book '68]

$$\mathbb{E}[P_e(t, d)] \leq A 2^{-E_J(B, R)nd}$$

- A is finite for $B < R_0$
- $E_J(B, R)$ properties:
 - $\frac{1}{2}E_G(R) \leq E_J(B, R) < E_G(R)$
 - $E_J(B, R) \xrightarrow{B \rightarrow R_0} E_G(R)$, for $R < R_{\text{crit}}$
- Does not guarantee a good **specific code** w.h.p.



Sequential Decoding: Error Probability

Proof for general codes requires:

- **Pairwise independence:** Every two paths are independent (from divergence point)
- **Individual codeword distribution:** Entries within each codeword are i.i.d.

Sequential Decoding: Error Probability

Use the following affine time-invariant ensemble:

$$\begin{bmatrix} \mathbf{c}_1 \\ \mathbf{c}_2 \\ \vdots \\ \mathbf{c}_t \\ \vdots \end{bmatrix} = \begin{bmatrix} \mathbf{G}_1 & \mathbf{0} & \cdots & \cdots & \cdots \\ \mathbf{G}_2 & \mathbf{G}_1 & \mathbf{0} & \cdots & \cdots \\ \vdots & \vdots & \ddots & \ddots & \cdots \\ \mathbf{G}_t & \mathbf{G}_{t-1} & \cdots & \mathbf{G}_1 & \mathbf{0} \\ \vdots & \vdots & \vdots & \vdots & \ddots \end{bmatrix} \begin{bmatrix} \mathbf{b}_1 \\ \mathbf{b}_2 \\ \vdots \\ \mathbf{b}_t \\ \vdots \end{bmatrix} + \begin{bmatrix} \mathbf{v}_1 \\ \mathbf{v}_2 \\ \vdots \\ \mathbf{v}_t \\ \vdots \end{bmatrix}$$

- Entries of $\{\mathbf{G}_t\}$, $\{\mathbf{b}_t\}$ and $\{\mathbf{v}_t\}$ are i.i.d. uniform
- $\{\mathbf{v}_t\}$ random translation vectors

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- Entries of $\{\mathbf{G}_t\}$, $\{\mathbf{b}_t\}$ and $\{\mathbf{v}_t\}$ are i.i.d. uniform
- $\{\mathbf{v}_t\}$ random translation vectors
- i.i.d. uniformity of $\{\mathbf{G}_t\} \Rightarrow$ pairwise independence ✓

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- $\{\mathbf{v}_t\}$ random translation vectors
- i.i.d. uniformity of $\{\mathbf{G}_t\} \Rightarrow$ pairwise independence ✓
- i.i.d. uniformity of $\{\mathbf{v}_t\} \Rightarrow$ i.i.d. uniformity of each codword ✓

Sequential Decoding: Complexity

- W_t – Number of branch computations at time t
- W_t is a random variable

Cutoff rate [Arikan IT'88]

For any “good” code (general or LTI), $\mathbb{E}[W_t]$ is unbounded for $R > R_0$.

Pareto distribution of W_t [Gallager, Zigangirov, Viterbi–Omura, ...]

$$\Pr(W_t \geq m) \leq Am^{-\rho}$$

- For $B, R < R_0$ and $R < \frac{B+R_0}{2\rho}$, $\rho \in (0, 1]$:
Tight for **general** and **LTI** codes $\Rightarrow \mathbb{E}[W_t] < \infty$ for $R < R_0$
- For $\rho > 1$, $R = E_0(\rho)/\rho$:
 - Tight for **general** codes
 - Widely conjectured to be true for **LTI** codes
- **Heavy tailed even if expectation is finite!**

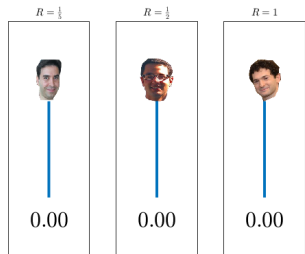
Simulation: Cart–Stick over BSC(0.01)

$$n = 20$$

$$k = 4 \quad 10 \quad 20$$

$$R = \frac{1}{5} \quad \frac{1}{2} \quad 1$$

$$E = 0.5382 \quad 0.2382 \quad 0$$



- Cart–stick system model
[Franklin–Powell–Emami-Naeini Book]
- BSC(0.01)
- For this setting
[Sukhavasi–Hassibi ISIT'11]:
 $k_{\min} = 3, E_{\min} = 0.2052$