

# Error Exponents of Typical Random Trellis Codes

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# Typical Random Codes

Traditional random coding error exponents are defined as

$$E_r(R) = \lim_{n \rightarrow \infty} \left[ -\frac{\ln \mathbf{E} P_e(\mathcal{C}_n)}{n} \right].$$

We define **typical**-code error exponents as

$$E_{\text{typ}}(R) = \lim_{n \rightarrow \infty} \left[ -\frac{\mathbf{E} \ln P_e(\mathcal{C}_n)}{n} \right].$$

- By Jensen's inequality,  $E_{\text{typ}}(R) \geq E_r(R)$ .
- $E_r(R)$  – dominated by **bad** codes;  $E_{\text{typ}}(R)$  – dominated by **typical** codes.

Let  $\mathcal{G}_E = \{\mathcal{C}_n : P_e(\mathcal{C}_n) \doteq e^{-nE}\}$ .

$$\overline{P_e(\mathcal{C}_n)} \doteq \sum_E P(\mathcal{G}_E) \cdot e^{-nE} \doteq P(\mathcal{G}_E^*) \cdot e^{-nE^*}.$$

Otoh,  $E_{\text{typ}}(R) = \sum_E P(\mathcal{G}_E) \cdot \mathbf{E} = E_0$ , where  $P[\mathcal{G}_{E_0}] \rightarrow 1$ .

# Motivation

- $E_{\text{typ}}(R)$  is never worse than  $E_r(R)$ .
- Code selected once and for all: no LLN to support  $\mathbf{EP}_e(\mathcal{C}_n)$ .
- Once selected, w.h.p.  $P_e(\mathcal{C}_n) \sim e^{-nE_0}$ , forever.
- Theoretical framework for random-like codes (Battail, 1995).
- Analogy: physics of disordered sys. – quenched vs. annealed average.

Q: With all these motivations, why wasn't it explored much more before?

A: Not so easy to analyze (also in physics) ....

# Related Work

- Barg & Forney ('02): i.i.d. random coding, BSC:

At low rates:  $E_{\text{typ}}(R) = E_{\text{ex}}(2R) + R$ .

- Nazari ('11); Nazari, Anastasopoulos & Pradhan ('14):

upper and lower bounds for the  $\alpha$ -decoder.

- Stat. phys. literature: Kabashima ('08), Mora & Riviore ('06), ...:

LDPC codes - replica analysis and cavity method.

- Battail ('95):

random-like codes.

- Merhav ('18): fixed-composition rand. coding, DMC, gen. likelihood dec.

Exact error exponent of the typical random code (TRC).

# Contributions

Deriving the error exponent of the typical random trellis code for:

- A general DMC (not just MBIOS).
- A general rational coding rate (not only  $1/n$ ).
- A general random selection of a time-varying code.

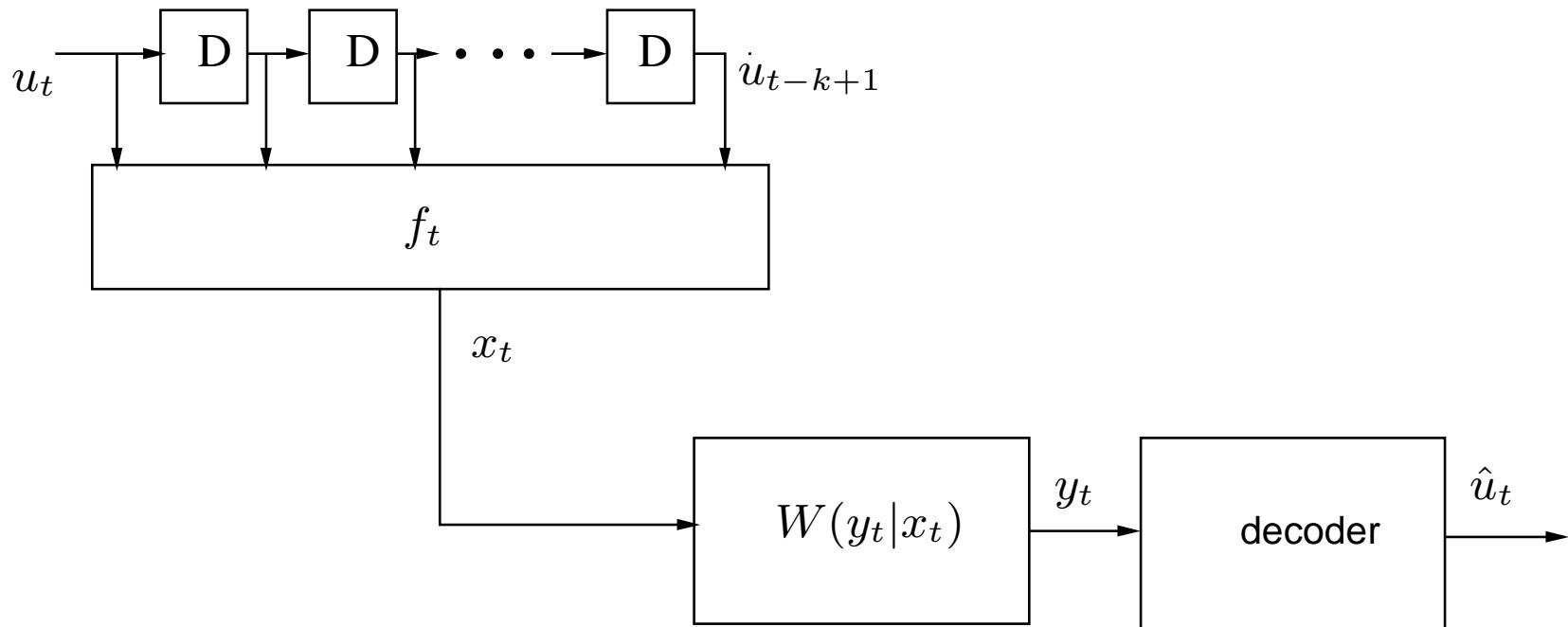
The analysis method (based on the MoT) provides insights on:

- Structure of the typical random trellis code.
- Dominant error events.

Additional extensions:

- Channels with finite input memory (ISI).
- Mismatched decoding metric.

# Problem Setting



- $\{u_t\}$  –  $m$ -vectors of purely random bits.
- $f_t : \{0, 1\}^{mk} \rightarrow \mathcal{X}^n$  randomly selected according to  $Q^n$ .
- Coding rate:  $R = m/n$ ; constraint length:  $K = mk$ .
- For convolutional codes,  $\{f_t\}$  are linear.
- $W = \text{DMC}$ .
- Asymptotic regime:  $k \rightarrow \infty$  while  $m$  and  $n$  are held fixed.

# Background

Traditional performance metric:  $E(R, Q) = \liminf_{K \rightarrow \infty} [-\log \mathbf{E} P_{\mathbf{e}}]/K$ .

$$E(R, Q) = \begin{cases} R_0(Q)/R & R < R_0(Q) \\ E_0(\rho, Q)/R & R > R_0(Q) \end{cases}$$

where  $\rho$  satisfies:  $\rho R = E_0(R, Q)$ .

For  $R > R_0(Q)$ :  $\exists$  matching converse.

For  $R < R_0(Q)$ : improvement by an expurgated bound,

$$R_0(Q)/R \rightarrow E_{\mathbf{cex}}(R, Q) \stackrel{\triangle}{=} E_{\mathbf{x}}(\rho, Q)/R$$

with  $\rho R = E_{\mathbf{x}}(\rho, Q)$ .

In [Viterbi & Odenwalder, 1969]: for **at least half** of the convolutional codes

$$P_{\mathbf{e}} \leq \left( \frac{2L}{1 - 2^{-\epsilon/\rho R}} \right)^\rho \cdot \exp\{-KE_{\mathbf{cex}}(R, Q)\}.$$

If  $2L$  is replaced by  $100L$ , the bound applies to 99% of the codes.

# Objectives

Studying the typical ensemble performance,

$$\mathcal{E}_{\text{trtc}}(R, Q) = \liminf_{K \rightarrow \infty} \frac{-\mathbf{E} \log P_{\mathbf{e}}}{K}$$

as well as  $\mathcal{E}_{\text{trcc}}(R, Q)$  defined similarly for convolutional codes.

- Deriving both “Csiszár–style” and “Gallager–style” expressions.
- Comparing to the random coding exponent and expurgated exponent.
- Comparing to typical random block codes of the same complexity.

# Main Result

For  $R < R_0(Q)$ :

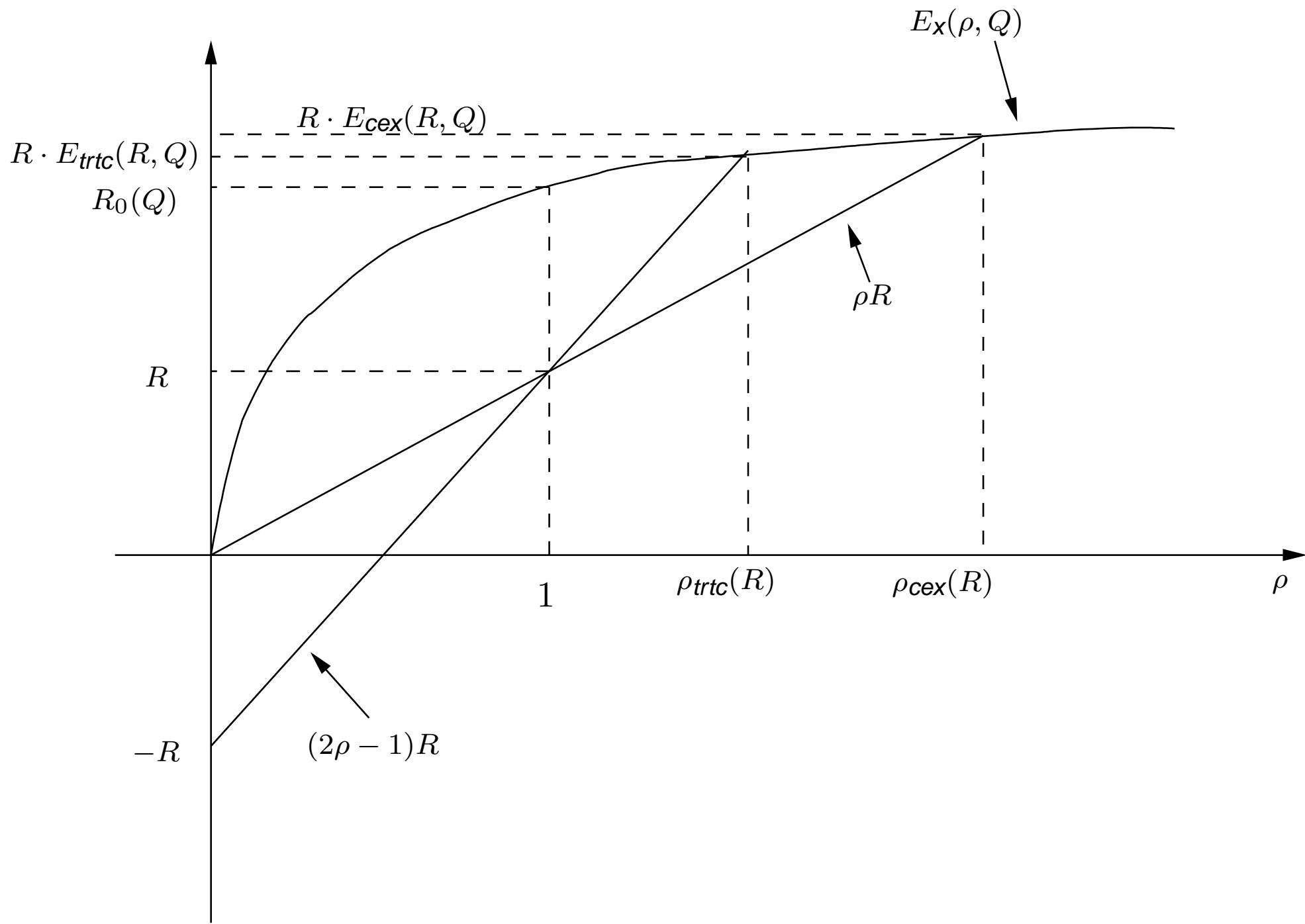
$$\mathcal{E}_{\text{trtc}}(R, Q) \geq E_{\text{trtc}}(R, Q) \triangleq \frac{E_{\mathbf{X}}(\rho, Q)}{R},$$

where  $\rho$  satisfies

$$\frac{E_{\mathbf{X}}(\rho, Q)}{2\rho - 1} = R.$$

Also,

$$\mathcal{E}_{\text{trcc}}(R, Q) \geq E_{\text{cex}}(R, Q).$$



# Characterizing the Typical Codes

The probability of error

$$P_{\mathbf{e}}(\mathcal{C}) \leq \sum_{\ell \geq 1} 2^{-m\ell} \sum_{\{P_{XX'}\}} N_{\ell}(P_{XX'}) \exp\{-n(k + \ell)\Delta(P_{XX'})\}.$$

For typical codes, all  $\{N_{\ell}(P_{XX'})\}$  that have a small expectation, **vanish simultaneously** w.h.p.

This amounts to the condition

$$2\ell R < (k + \ell)D(P_{XX'} \| Q \times Q).$$

Joint types that are “too far” from  $Q \times Q$  are not populated.

For populated joint types,

$$N_{\ell}(P_{XX'}) \sim \exp\{m[2\ell - (k + \ell)D((P_{XX'} \| Q \times Q)/R]\}.$$

# An Alternative Expression

$$E_{\text{trtc}}(R, Q) = \inf_{\hat{R} < R} \inf_{\{P_{XX'} : D(P_{XX'} \| Q \times Q) < 2\hat{R}\}} \frac{\mathbf{E}_P d_{\mathbf{B}}(X, X') + \hat{R}}{R - \hat{R}}.$$

Dominant error events:

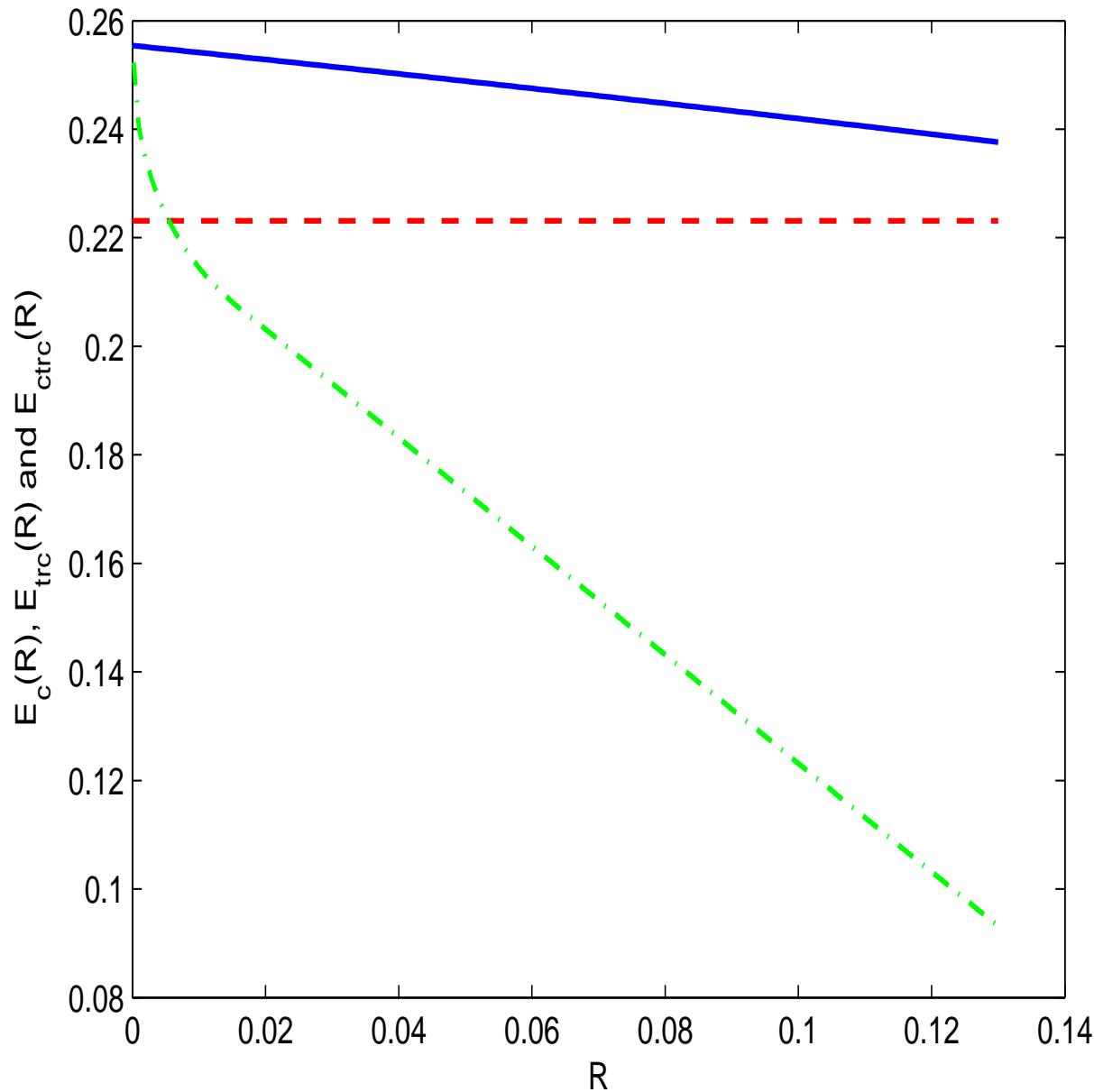
A sub-exponential number of paths with joint type,

$$P_{XX'}(x, x') = \frac{Q(x)Q(x') \exp\{-d_{\mathbf{B}}(x, x')/\rho\}}{Z}$$

and critical length of

$$k + \ell = \frac{kR}{2R - D(P_{XX'} \| Q \times Q)}.$$

# A Numerical Example: BSC with $p = 0.1$



## Two Words Regarding an Extension

The paper contains also an extension to a channel with mismatch and input memory (ISI),

$$W(\mathbf{y}|\mathbf{x}) = \prod_t W(y_t|x_t, x_{t-1}).$$

$E_{\mathbf{x}}(\rho, Q)$  is now replaced by  $-\rho \log \lambda$ , where  $\lambda$  is the **Perron-Frobenius eigenvalue** of a certain matrix that depends on the  $Q$  and on the channel single-letterly (details – in the paper).