

Sequential Decoding of Convolutional Codes for Synchronization Errors

Anisha Banerjee

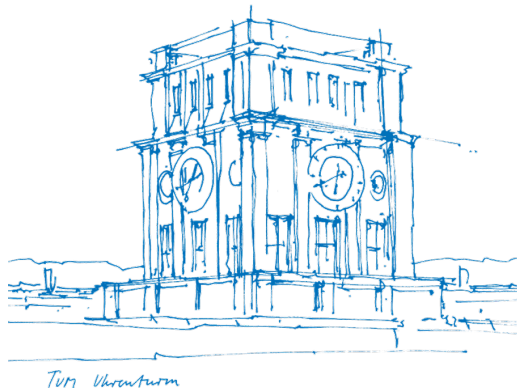
Joint work with

Andreas Lenz, Antonia Wachter-Zeh

Technical University of Munich

Institute for Communications Engineering

November 8, 2022



Outline

Introduction

Sequential Decoding

Modifications for IDS Channels

Results & Conclusion

Introduction

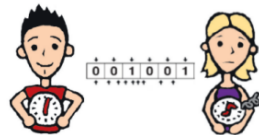
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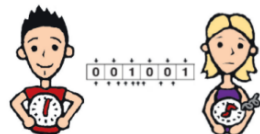
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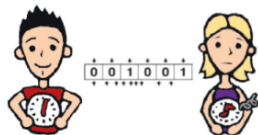
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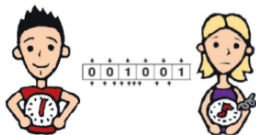
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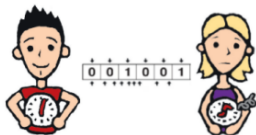
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- May use **convolutional codes** for correction.
- **Problem:** High complexity of Viterbi & MAP decoders.
- **Solution:** Use **sequential decoders**!
 - ▶ Only examines 'promising' codewords.
 - ▶ Complexity independent of memory.



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Channel Model

- Permits insertions, deletions and substitutions at random positions.
→ Denote as '**IDS channel**'.

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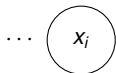
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 - ▶ Let channel input be $\mathbf{x} = (x_1, \dots, x_T)$.
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Overview

Prior work

- [MT10] & [BF15] adapted Viterbi & MAP decoders to accommodate indels.
- [Gal61] & [MT02] investigated sequential decoding for IDS channels.

Our contributions

- New **decoding metric** for Fano's sequential decoder.
- Determination of '**cutoff rate**'.
→ Beyond this code rate, sequential decoder is computationally impractical.

[MT10] M. F. Mansour and A. H. Tewfik, "Convolutional decoding in the presence of synchronization errors," *IEEE Journal on Selected Areas in Communications*, vol. 28, no. 2, pp. 218–227, Feb. 2010

[BF15] V. Buttigieg and N. Farrugia, "Improved bit error rate performance of convolutional codes with synchronization errors," in *Proc. Int. Conf. Comm.*, London, Jun. 2015, pp. 4077–4082

[Gal61] R. G. Gallager, "Sequential decoding for binary channel with noise and synchronization errors," *Lincoln Lab Group*, Arlington, VA, USA, Tech. Rep., Sep. 1961

[MT02] M. F. Mansour and A. H. Tewfik, "Convolutional codes for channels with substitutions, insertions, and deletions," in *Proc. Global Commun. Conf.*, vol. 2, Taipei, Taiwan: IEEE, 2002, pp. 1051–1055

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Example

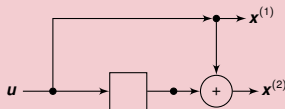


Figure: A $(2, 1, 1)$ binary convolutional encoder

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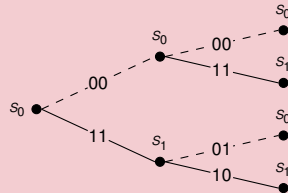


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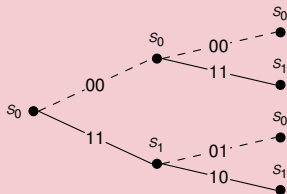


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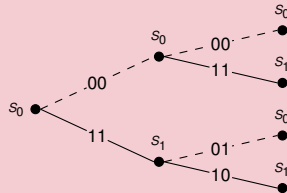


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Example

Say we receive $\mathbf{y} = 1010$ over a BSC with $P_s = 0.04$.

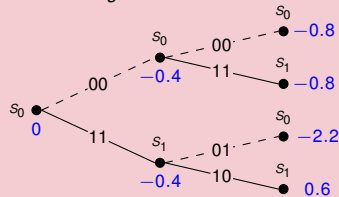
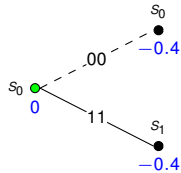


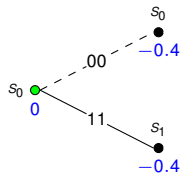
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Fano's Decoder



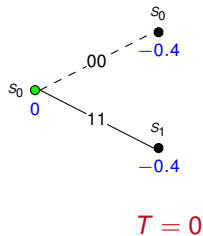
- **Decoder** starts at root; evaluates successors' metrics.

Fano's Decoder



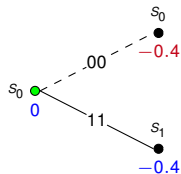
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Fano's Decoder



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- Initialize **threshold**.

Fano's Decoder



$$\Delta = 0.5$$

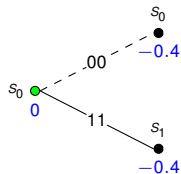
$$T = 0$$

$$\mu_s = -0.4$$

Algorithm

- Is **best successor metric** $\geq T$?

Fano's Decoder



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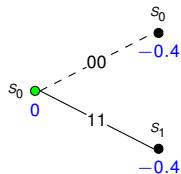
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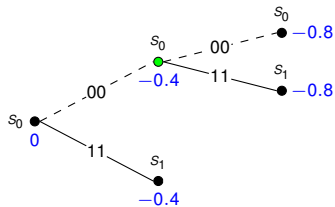
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Fano's Decoder



$$\Delta = 0.5$$

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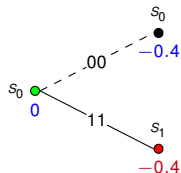
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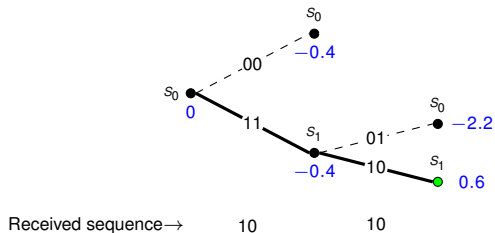
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Fano's Algorithm



Summary

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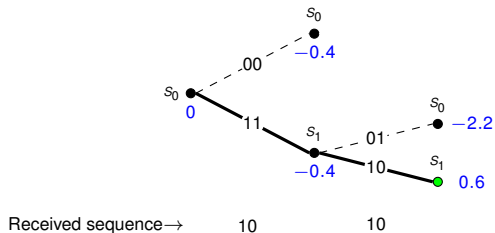
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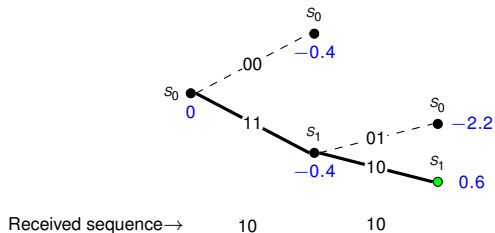
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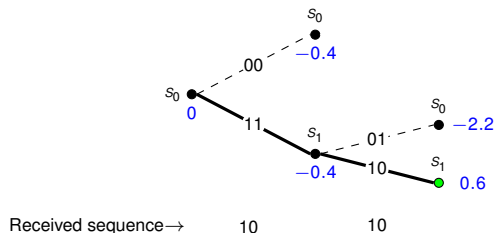
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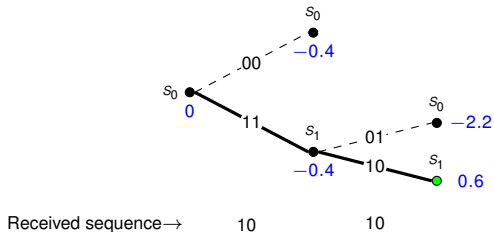
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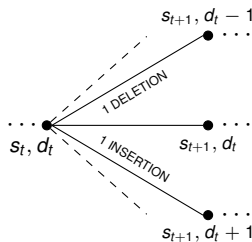


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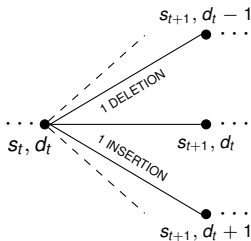


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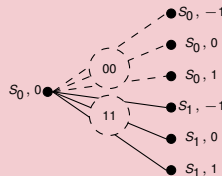


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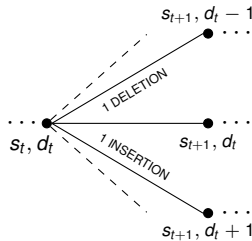


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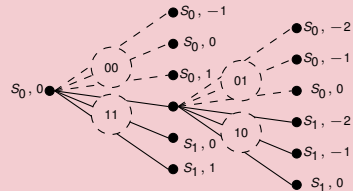


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- **Special case:** When no $P_i = P_d = 0$, ignore drift states.

$$\mu(v_t) = \log P(\mathbf{s}|\mathbf{y})$$

→ **Original Fano metric** for substitution-only channels!

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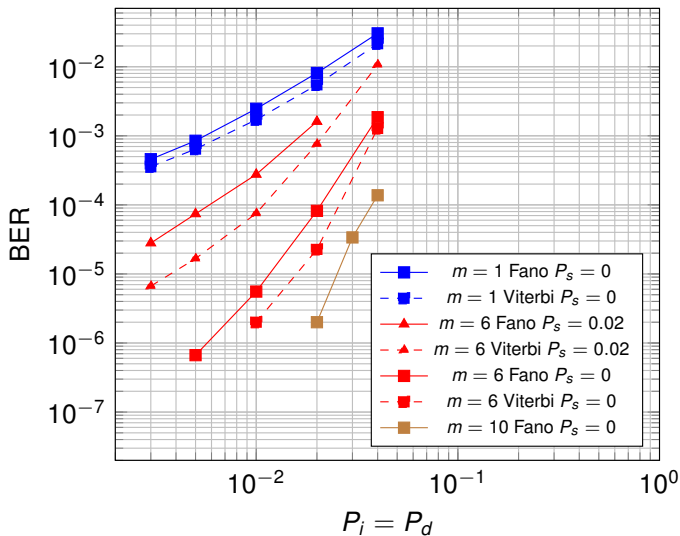
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Bit Error Rate

#Blocks=300, $c = 3$, $b = 1$, Terminated



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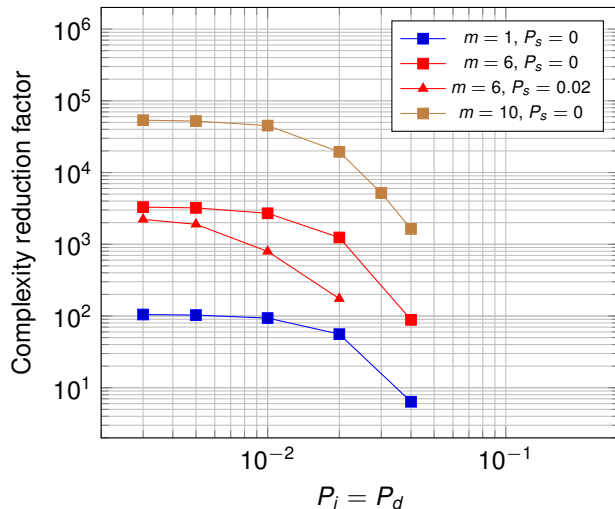
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- Define

$$\begin{aligned} \text{Complexity reduction factor} &= \frac{\text{Complexity of Viterbi}}{\text{Mean complexity of Fano's}} \\ &= \frac{\text{\#Nodes in trellis}}{\text{\#Mean steps forward}} \end{aligned}$$

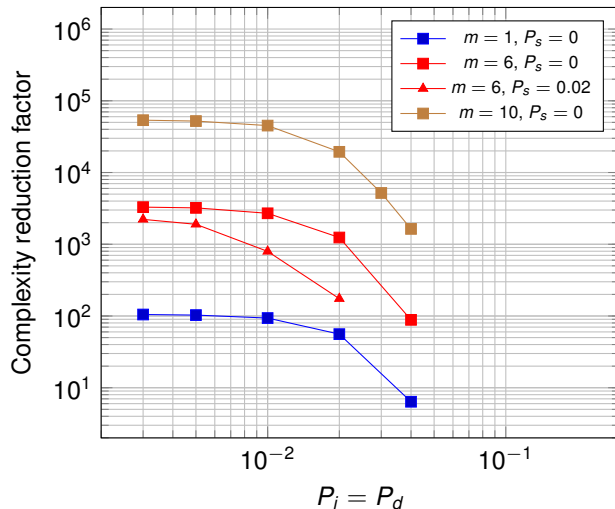
Simulated Complexity Comparison

#Blocks=300, c=3, b=1, Terminated



Simulated Complexity Comparison

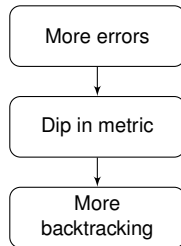
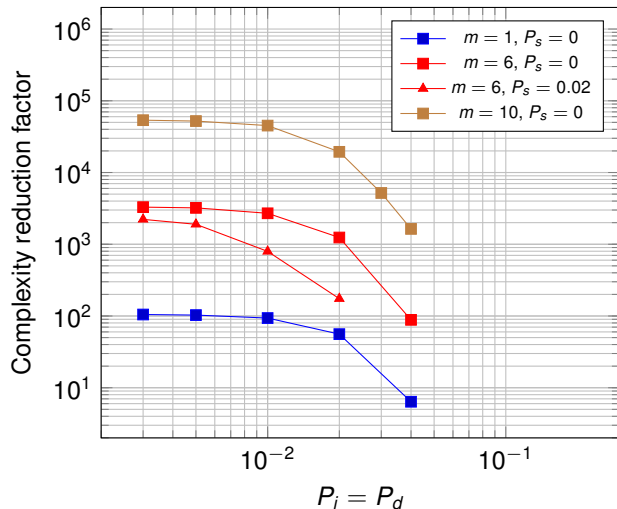
#Blocks=300, c=3, b=1, Terminated



More errors

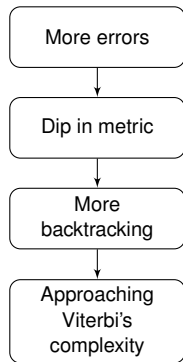
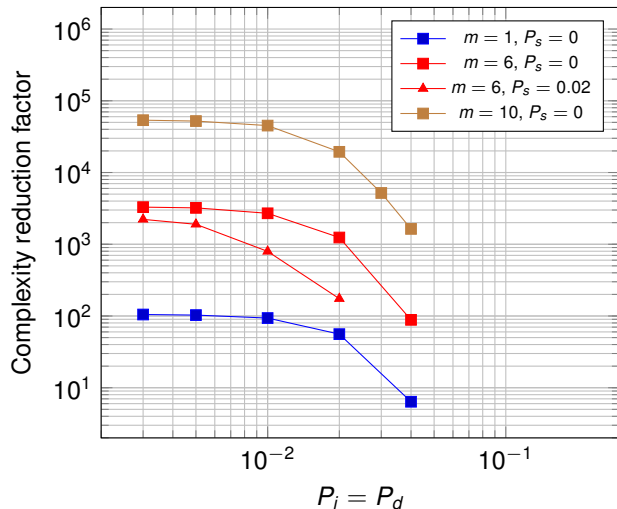
Simulated Complexity Comparison

#Blocks=300, c=3, b=1, Terminated



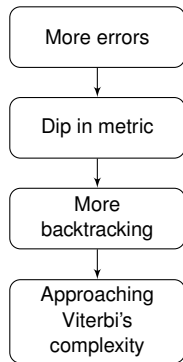
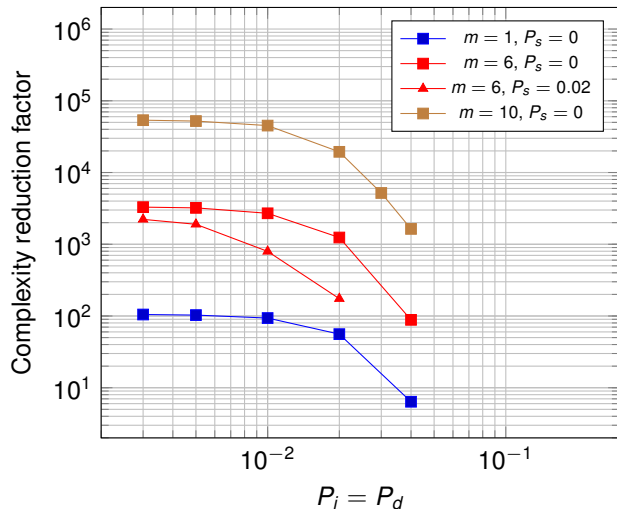
Simulated Complexity Comparison

#Blocks=300, c=3, b=1, Terminated



Simulated Complexity Comparison

#Blocks=300, c=3, b=1, Terminated



Question: When is Fano's algorithm no longer practical to use?

Computational Cutoff Rate

- Beyond cutoff rate R_0 , sequential decoder is computational impractical.

[JZ15] R. Johannesson and K. S. Zigangirov, "Sequential decoding," in *Fundamentals of Convolutional Coding*. Hoboken, New Jersey: John Wiley & Sons, Inc., 2015, pp. 425–484

Computational Cutoff Rate

- Beyond cutoff rate R_0 , sequential decoder is computational impractical.
- For **rates** $< R_0$,
 - ▶ Complexity of decoding one frame grows **linearly** with #blocks.

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Computational Cutoff Rate

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- For $\text{rates} < R_0$,
 - ▶ Complexity of decoding one frame grows **linearly** with #blocks.
- For $\text{rates} > R_0$,
 - ▶ Too much backtracking.
 - ▶ Complexity of decoding one frame grows **exponentially** with #blocks..

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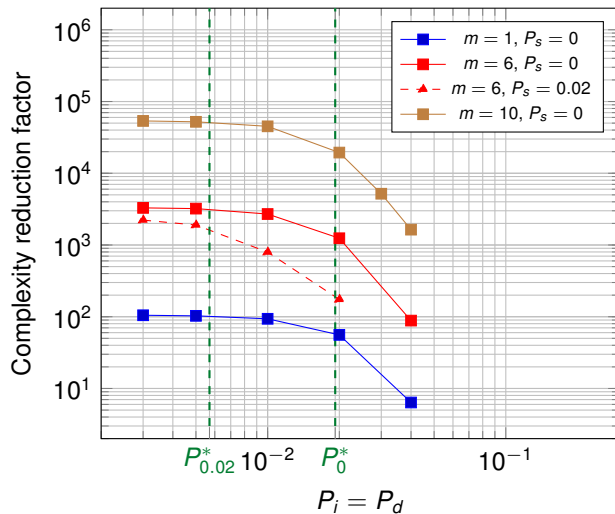
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- By extending methods in [JZ15], we **can compute this!**

[JZ15] R. Johannesson and K. S. Zigangirov, "Sequential decoding," in *Fundamentals of Convolutional Coding*. Hoboken, New Jersey: John Wiley & Sons, Inc., 2015, pp. 425–484

Simulated Complexity Comparison

#Blocks=300, c=3, b=1, Terminated



P_0^* & $P_{0.02}^*$ mark cutoff rate operation for code (3, 1) at $P_s = 0$ and $P_s = 0.02$ respectively.

Conclusion

Summary

- New **decoding metric** for Fano's algorithm in IDS channels.
- Determination of computational **cutoff rate**.

Future work

- Error probability analysis.
- Branching process techniques for complexity analysis.

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Thank you!

References

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