

Quantized Polar Code Decoders: Analysis and Design

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Thank you: Rüdiger Urbanke^{EPFL}, Gerhard Kramer^{TUM}

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Lower complexity (e.g., IoT)
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→ Cheaper device production

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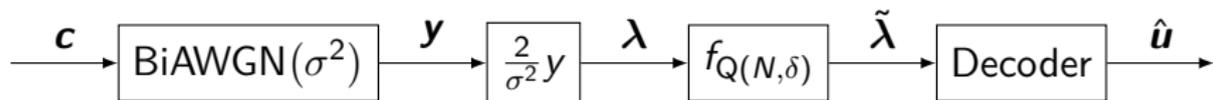
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- Sizable gains, particularly for low code rates

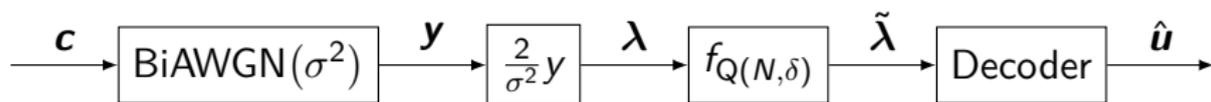


Preliminaries

System Model

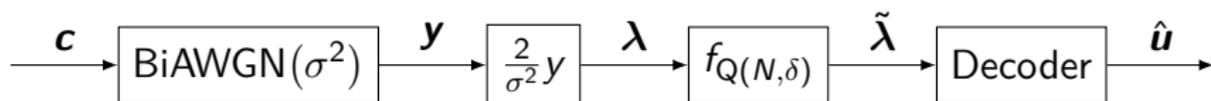


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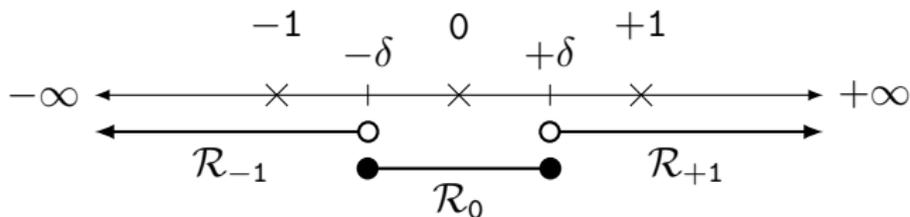


Uniform quantization $f_{Q(3,\delta)}$: $\mathcal{L}_3 \triangleq \{-1, 0, +1\} \subseteq \mathbb{Z}$

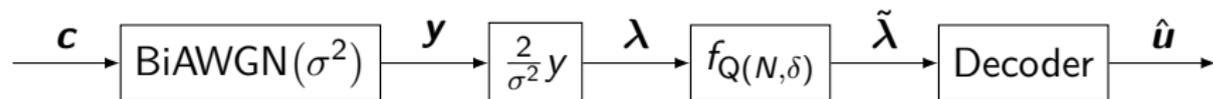
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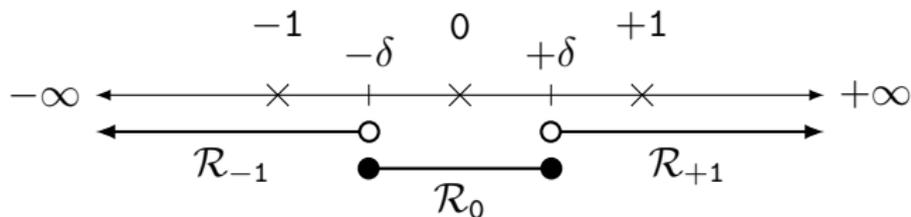
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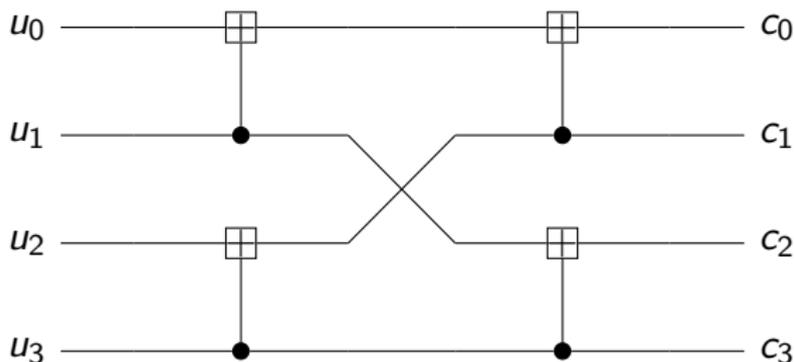
Other quantization: $\mathcal{L}_7 \triangleq \{0, \pm 1, \pm 2, \pm 3\} \subseteq \mathbb{Z}$ $\mathcal{L}_\infty \triangleq \mathbb{R}$

PC Basics & SC Decoding [Stolte 2002] [Arıkan 2009]

$$\mathbf{c} = \mathbf{G}\mathbf{u} \quad \mathbf{G} = \mathbf{F}^{\otimes m} \mathbf{P}_m^{(\text{bitrev})} \quad \mathbf{F} = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \quad n = 2^m$$

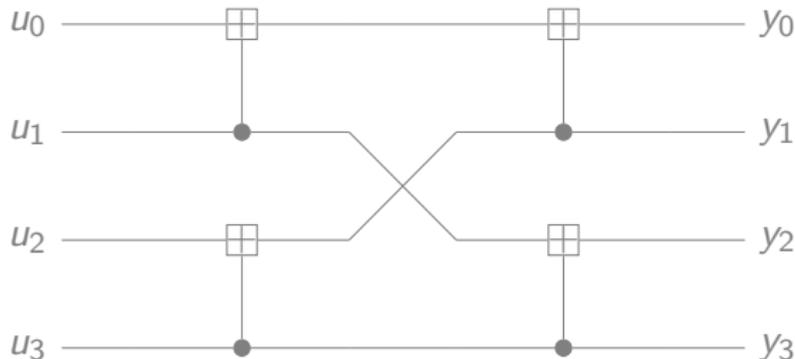
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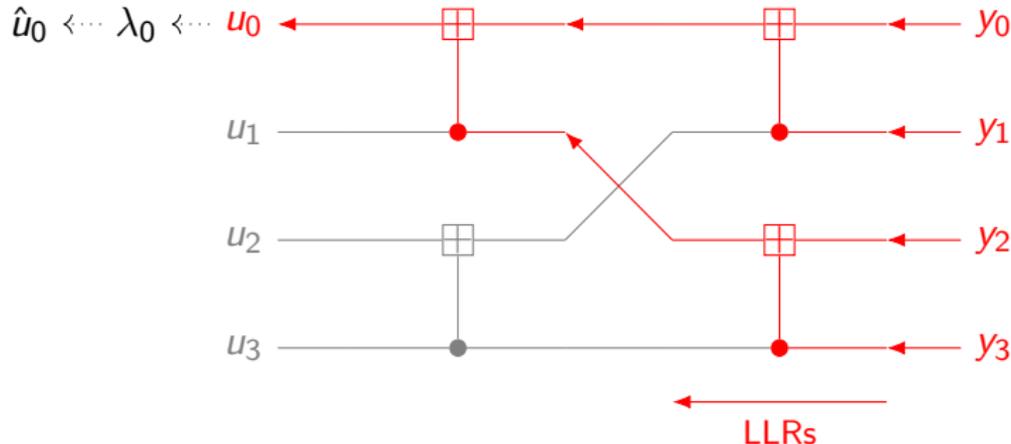
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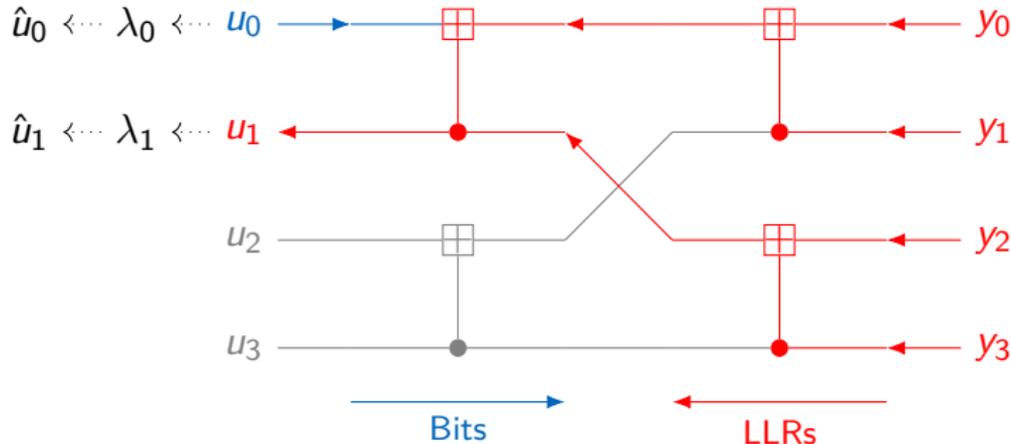
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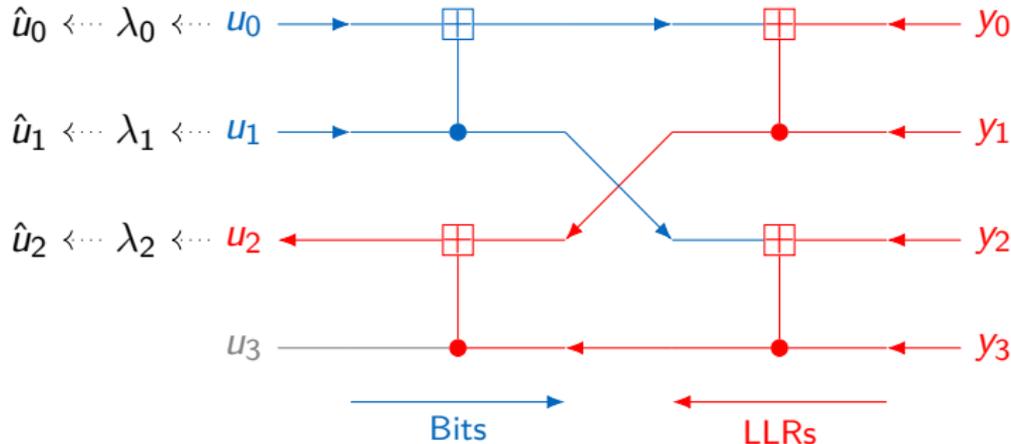
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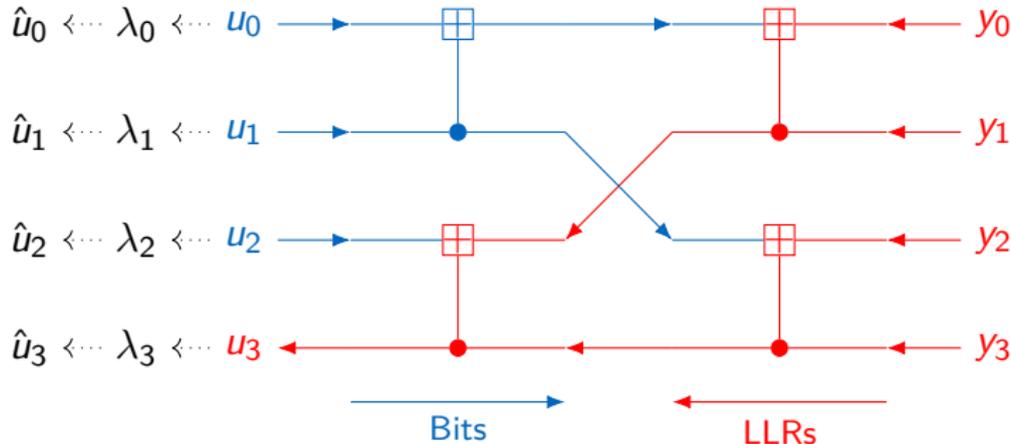
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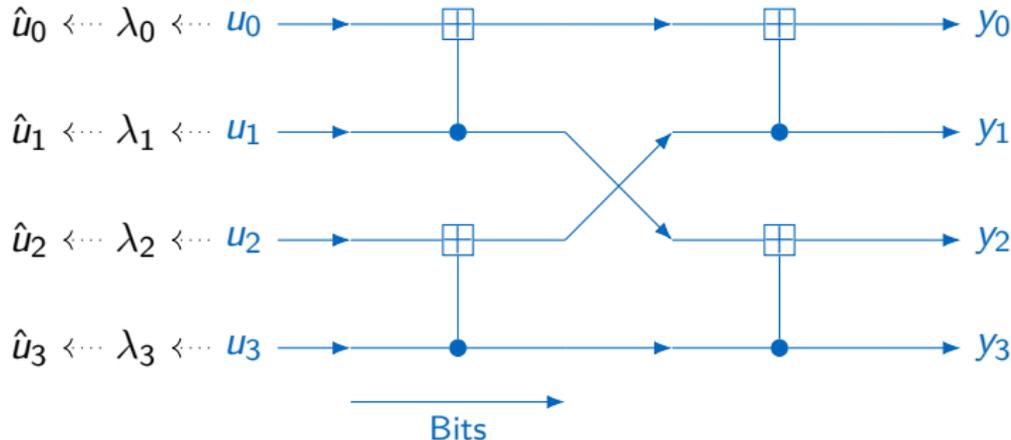
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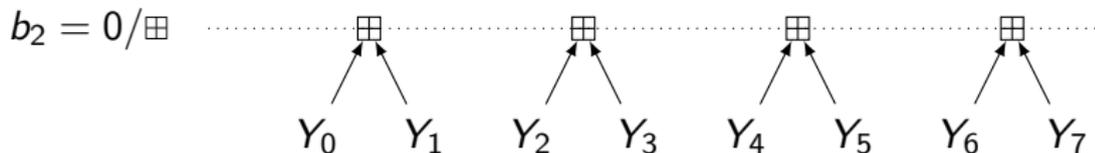
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Y_0 Y_1 Y_2 Y_3 Y_4 Y_5 Y_6 Y_7

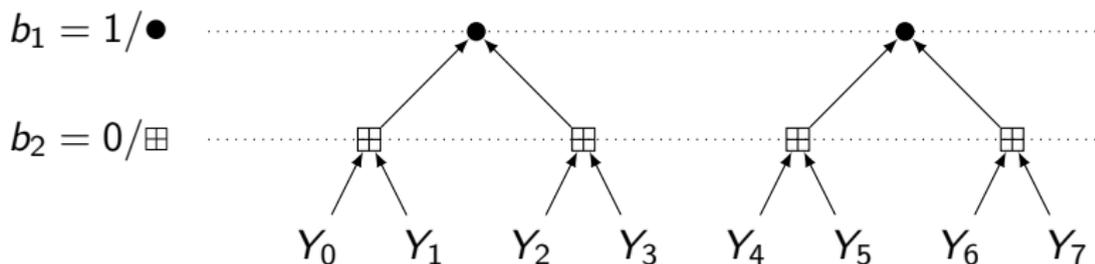
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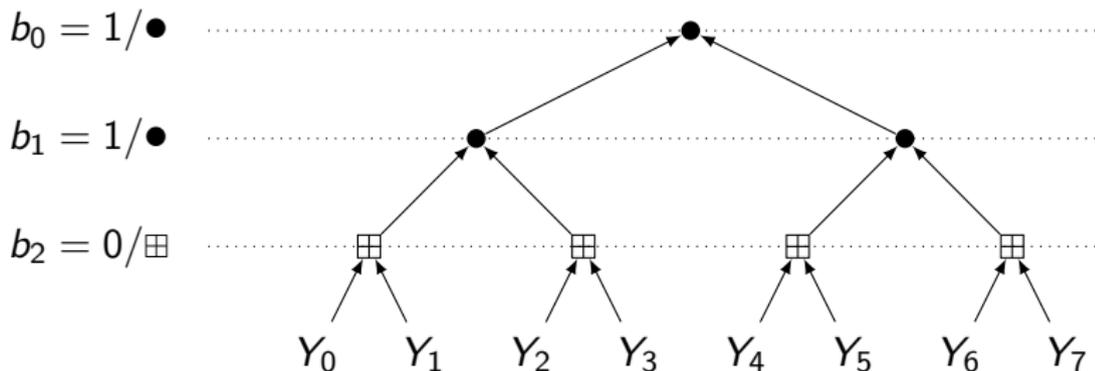
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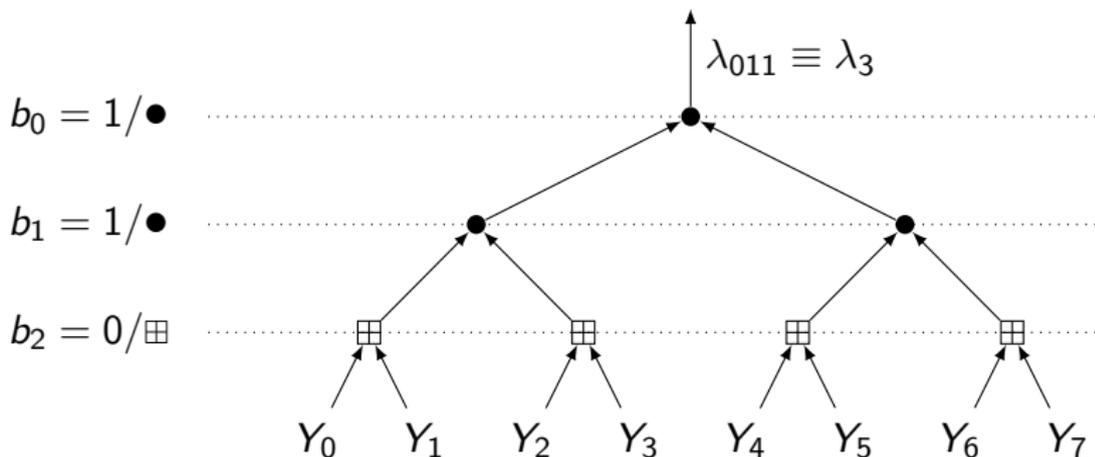
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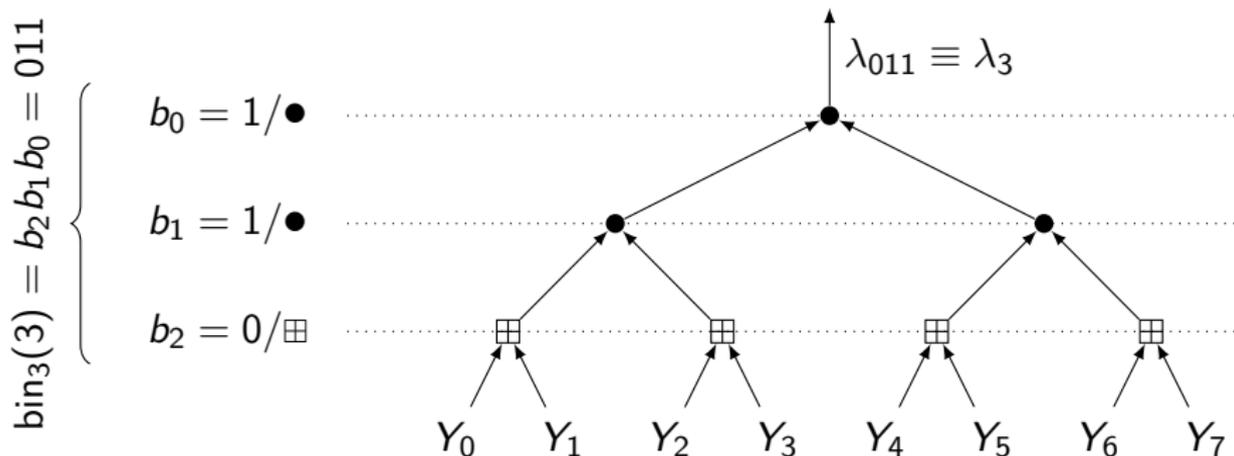
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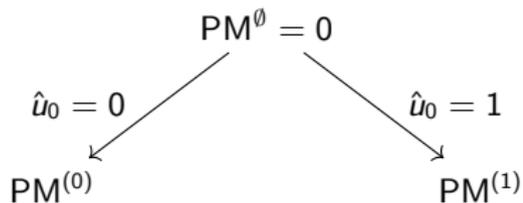
SC List Decoding [Tal and Vardy 2015] [Balatsoukas-Stimming et al. 2015]

List size $L = 2$:

$$PM^\emptyset = 0$$

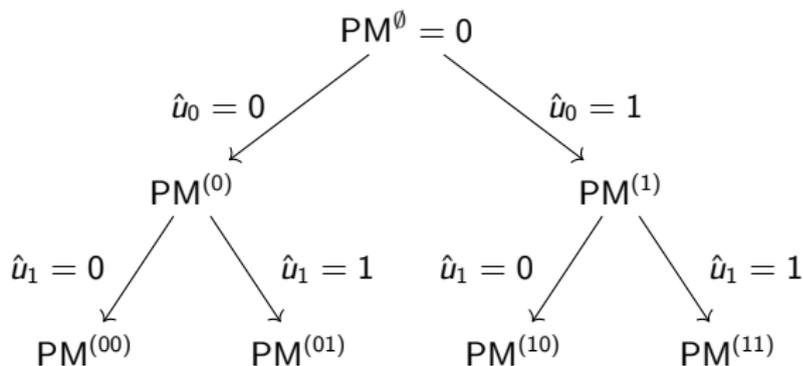
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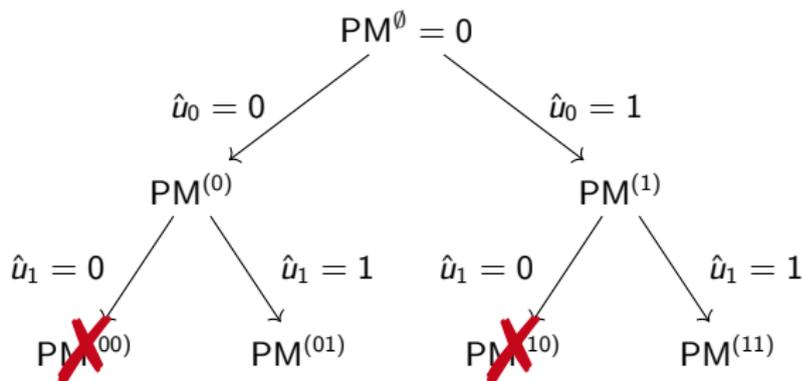
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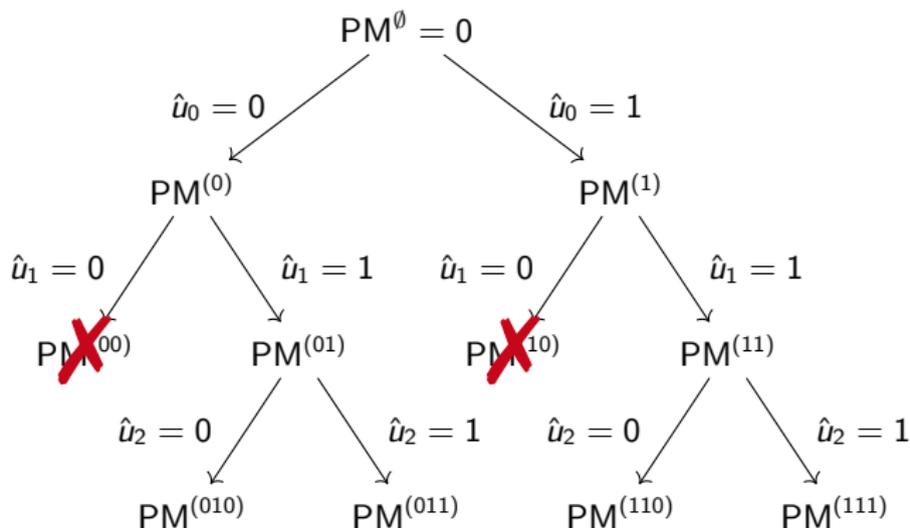
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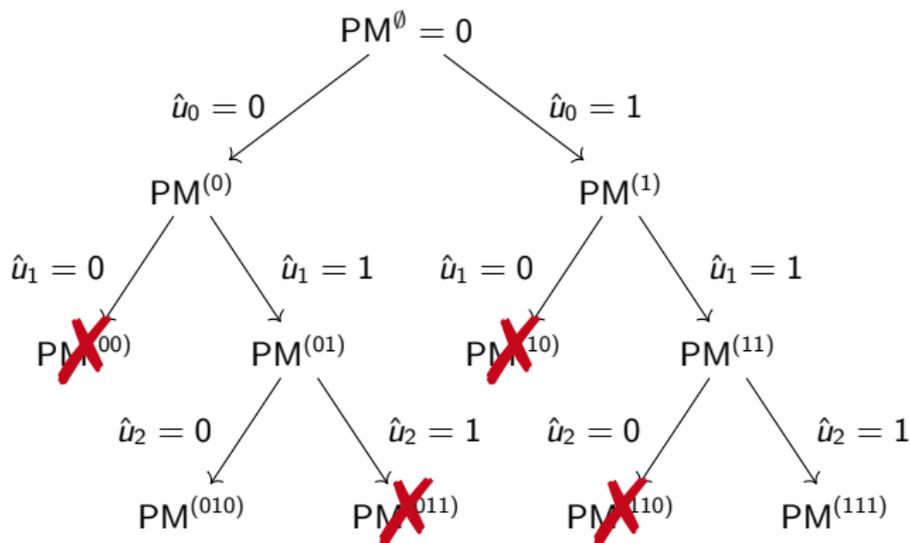
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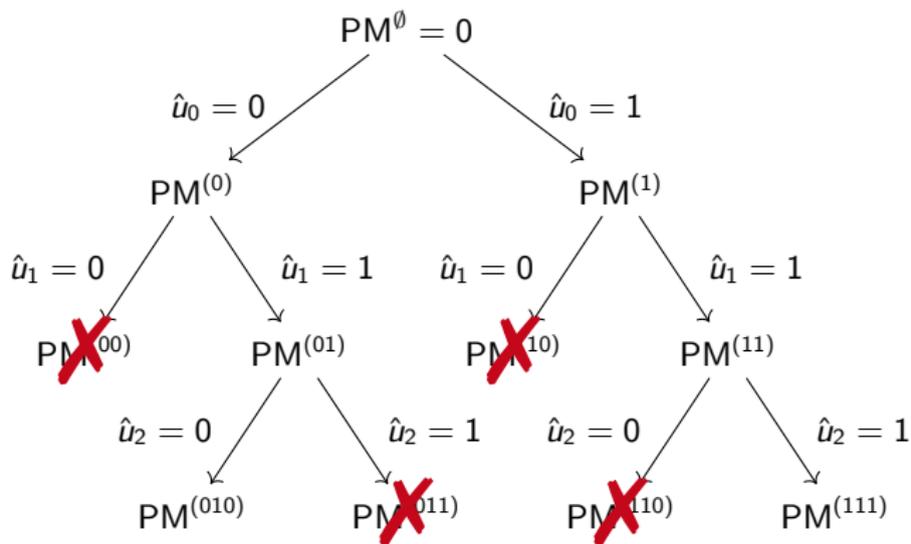
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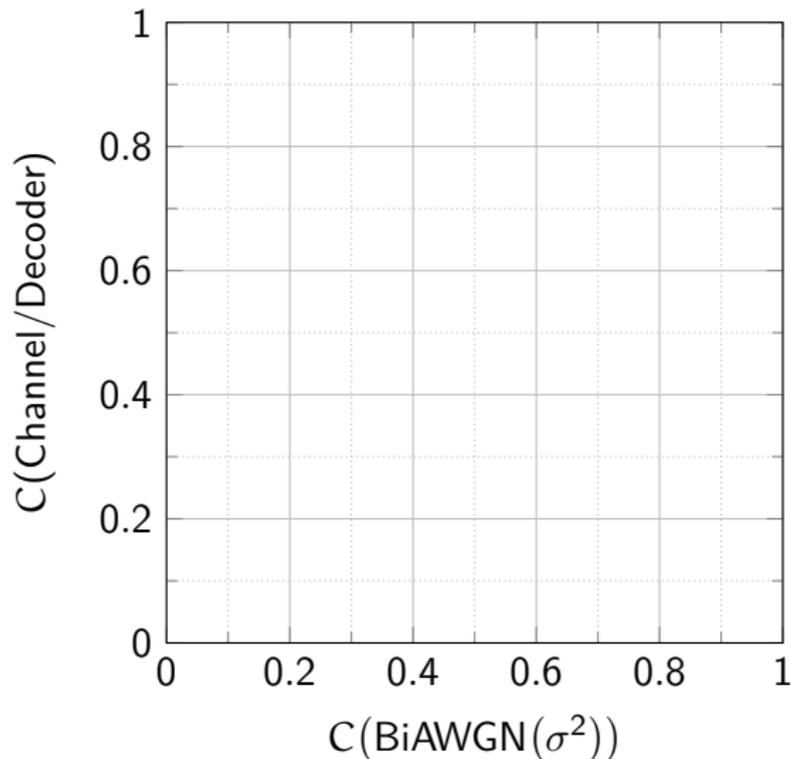


→ List $\left\{ (010, PM^{(010)}), (111, PM^{(111)}) \right\}$ of length $L = 2$

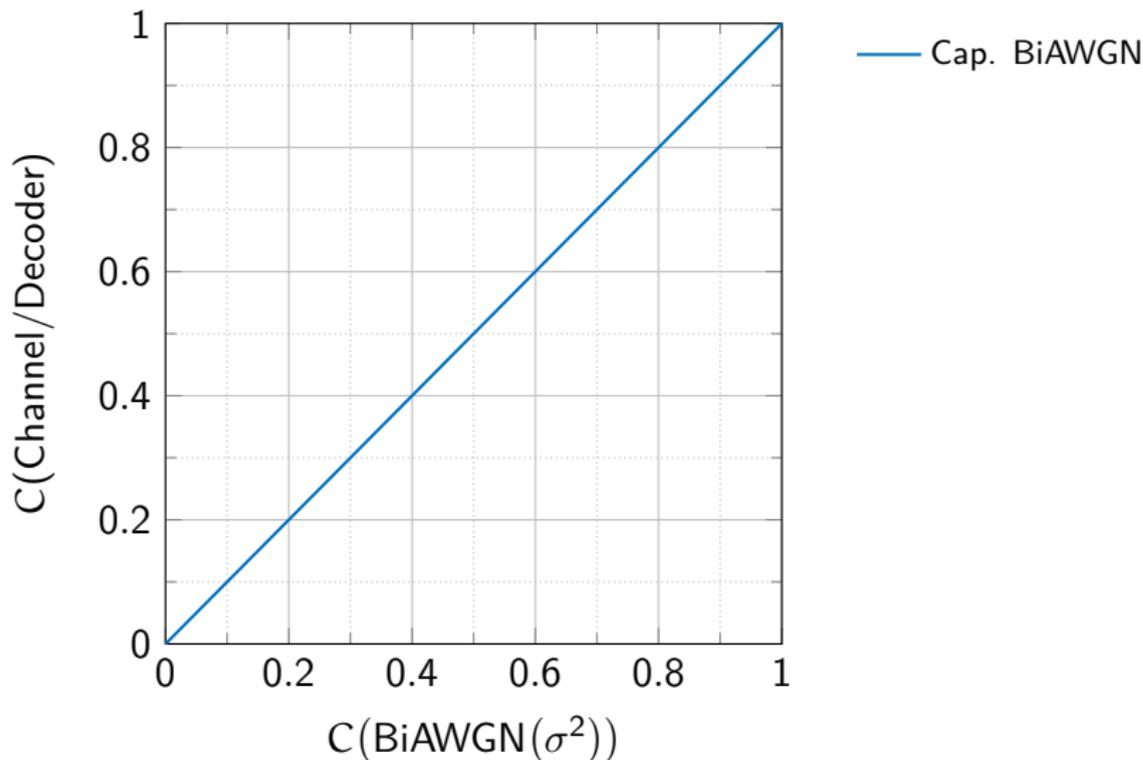


3Q Decoding

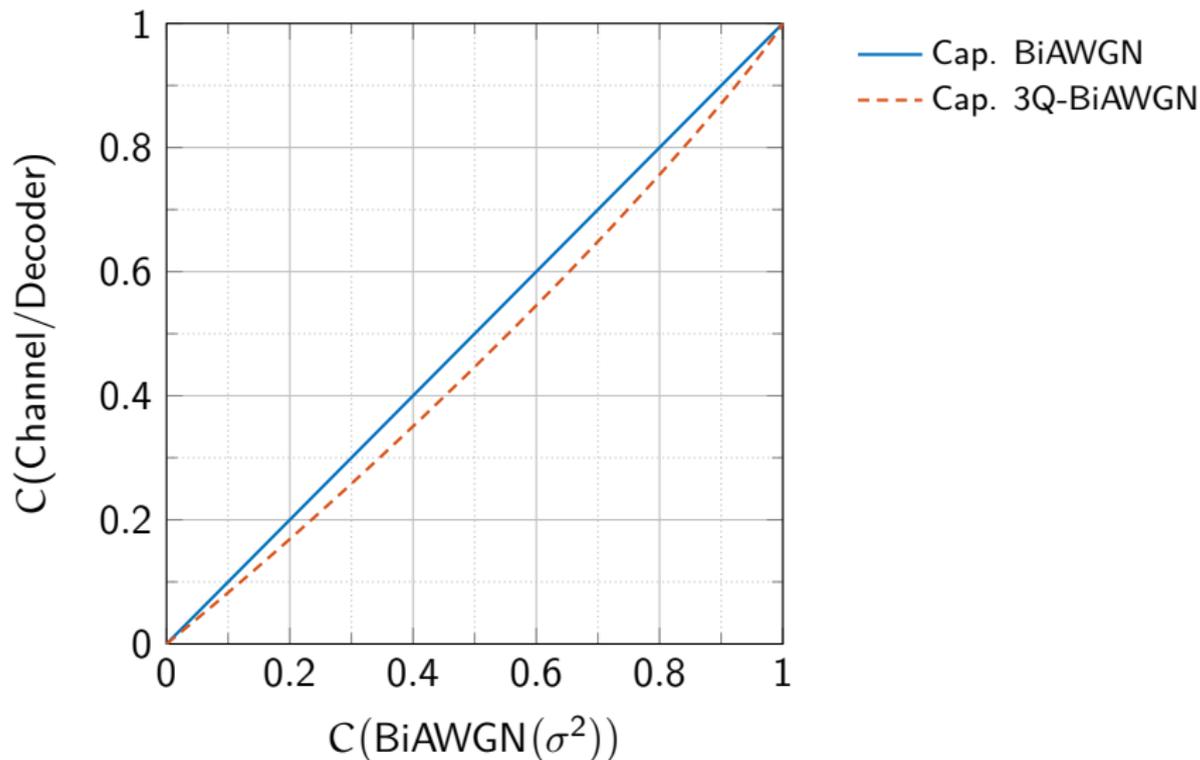
3Q-SC Decoding Capacity [Hassani and Urbanke 2012]



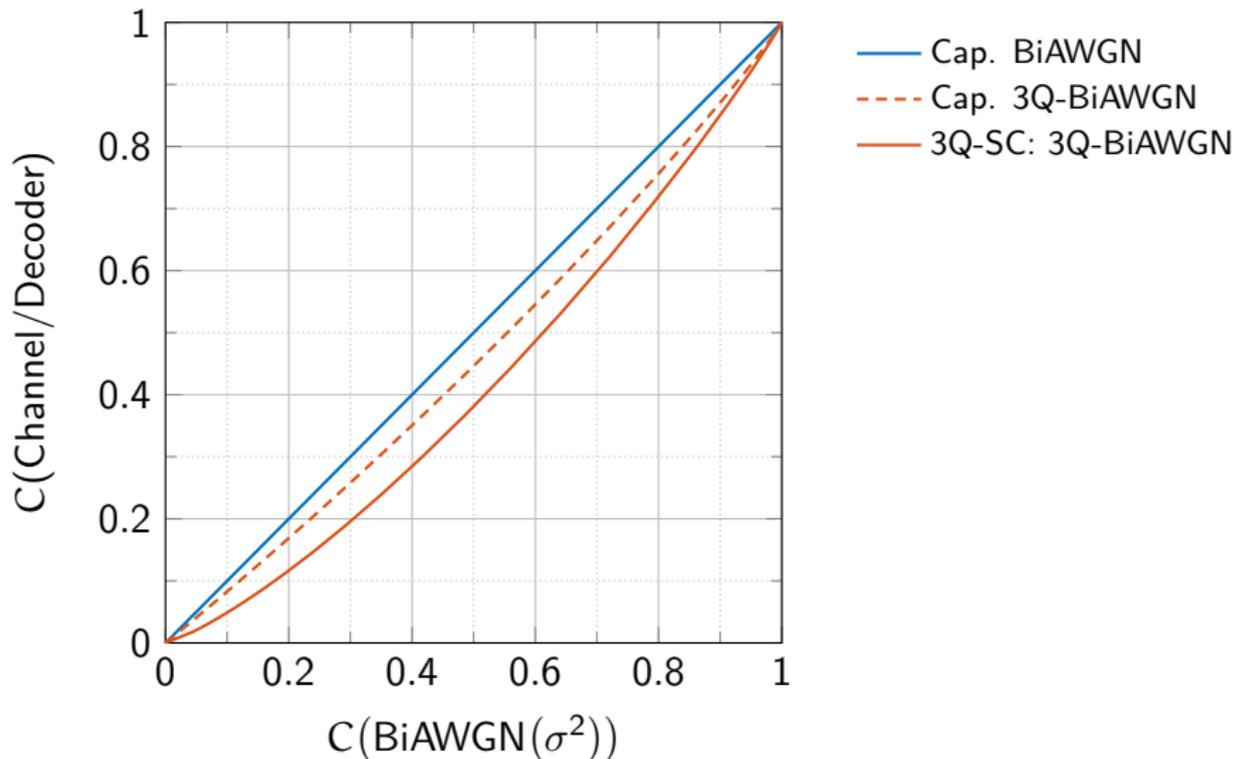
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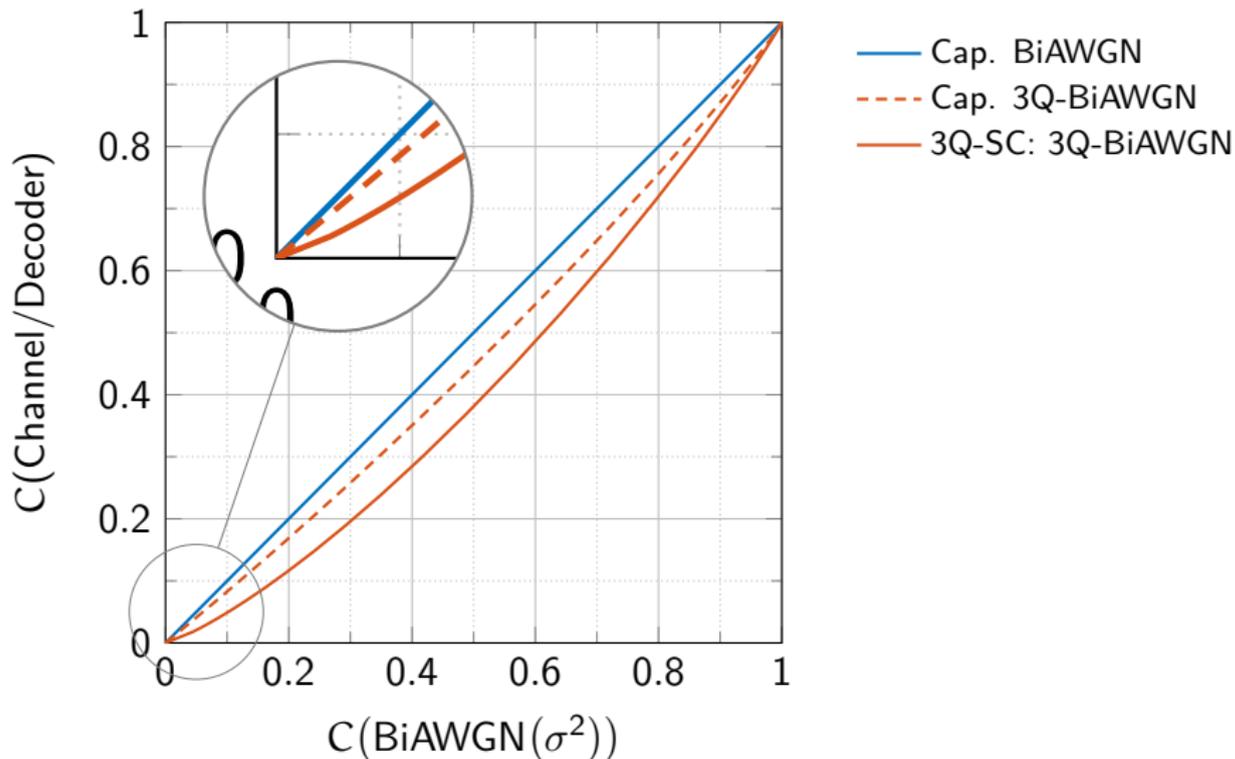
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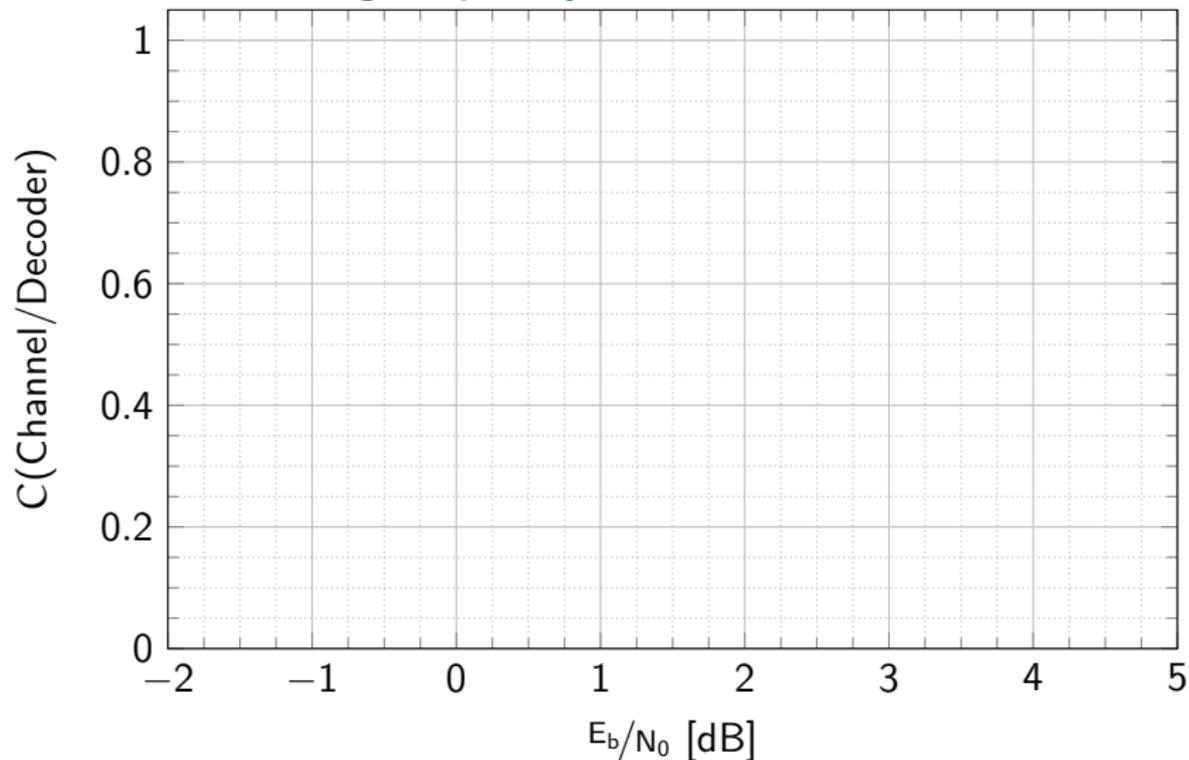
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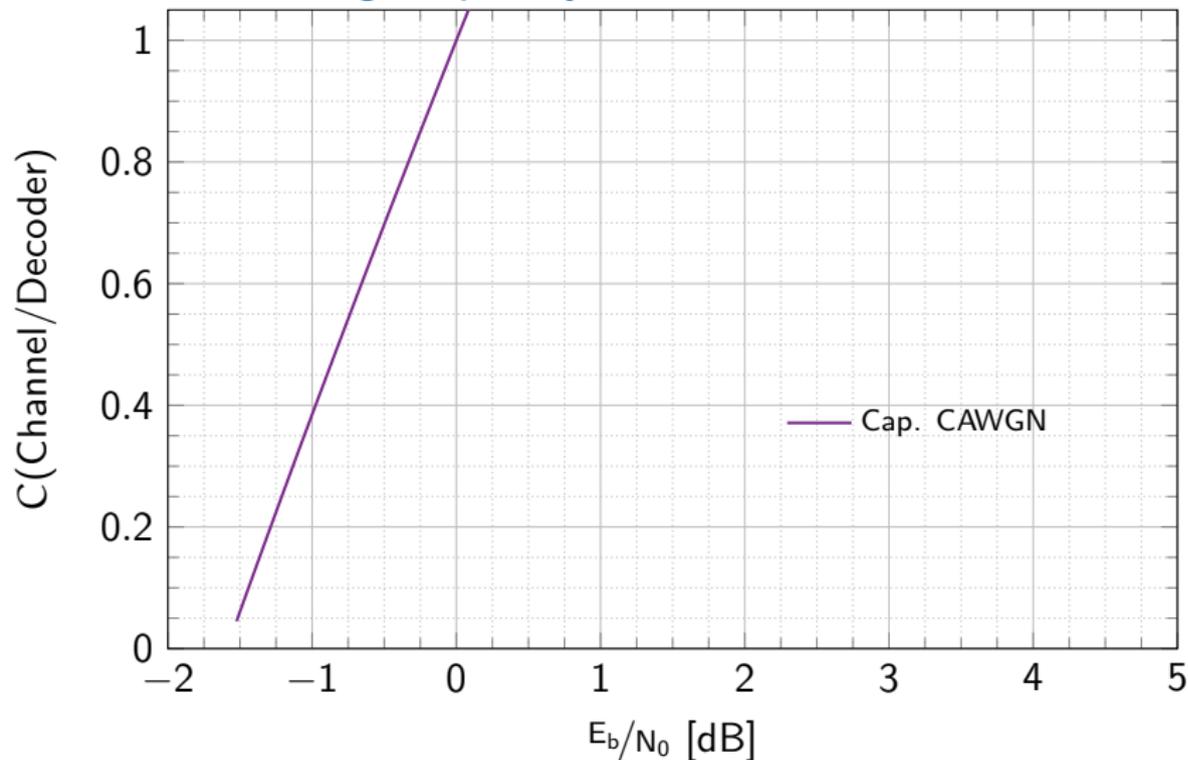
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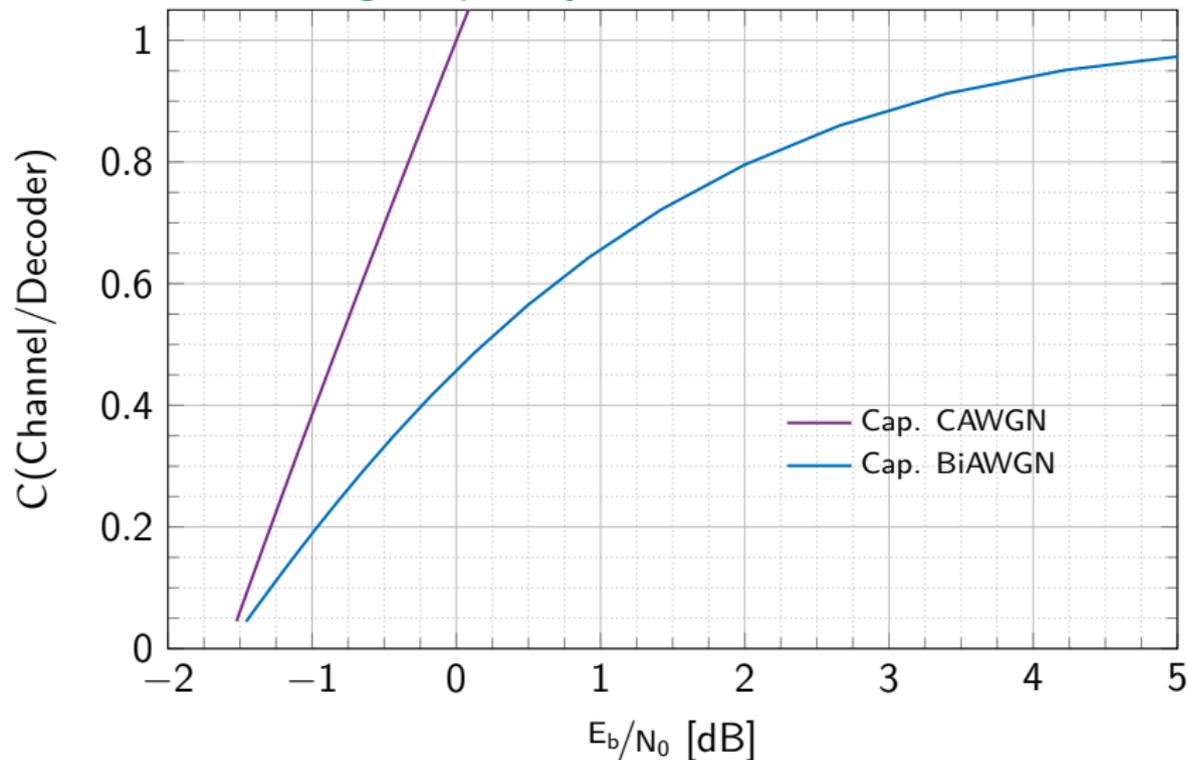
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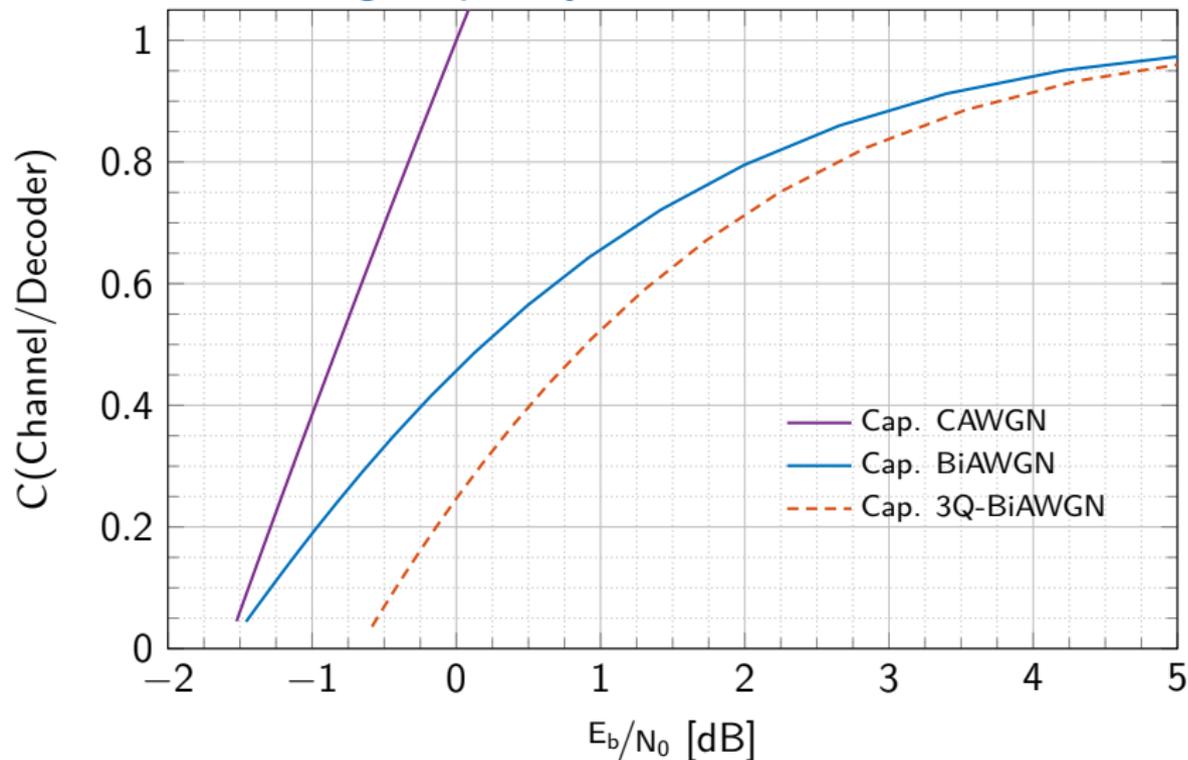
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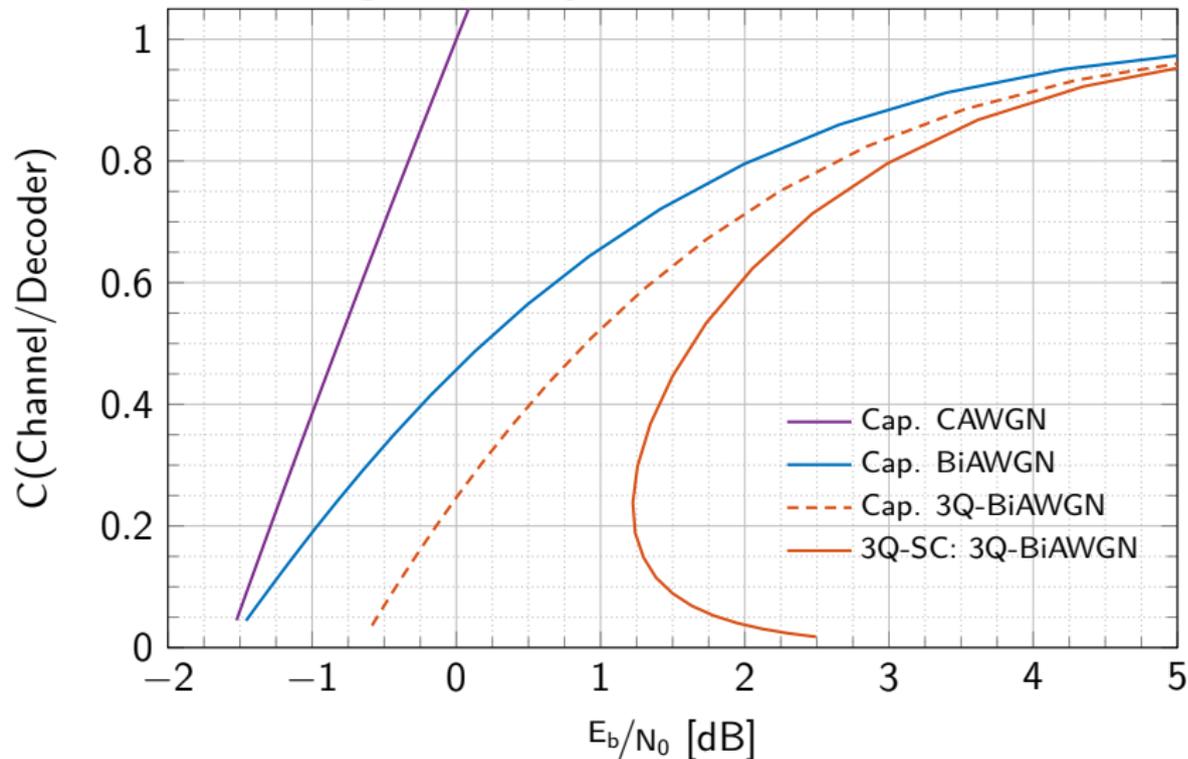
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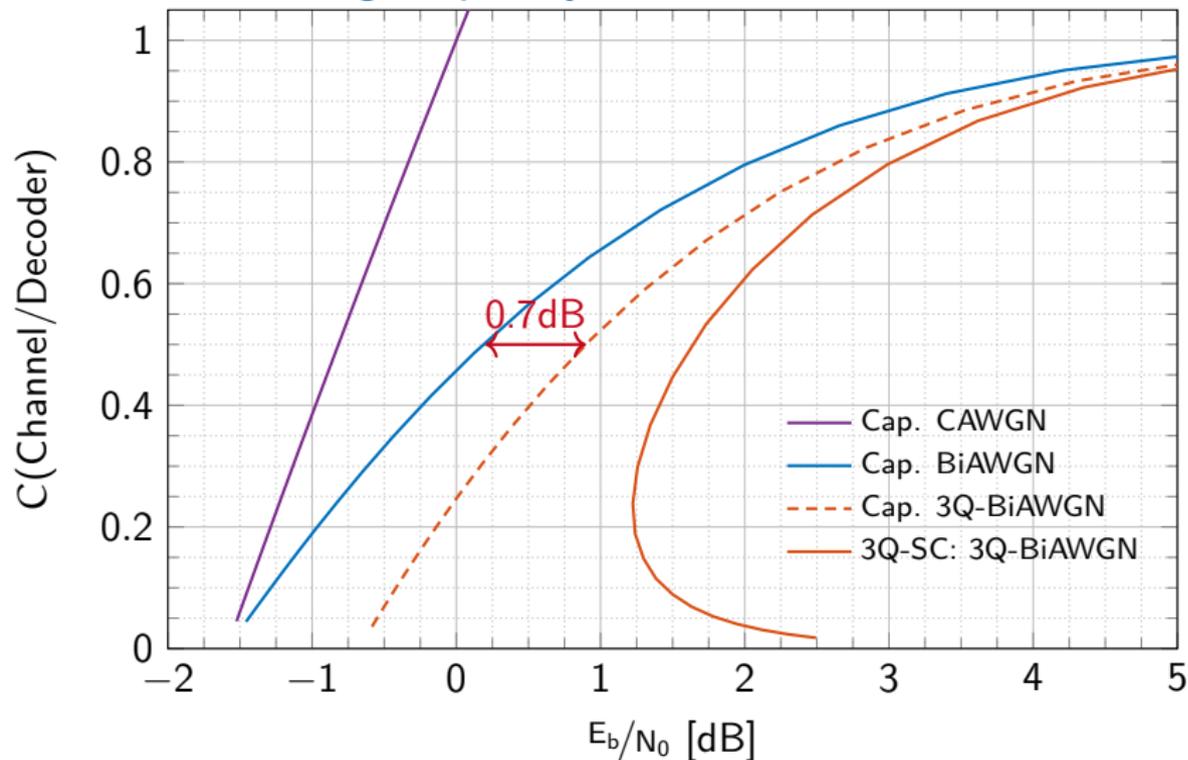
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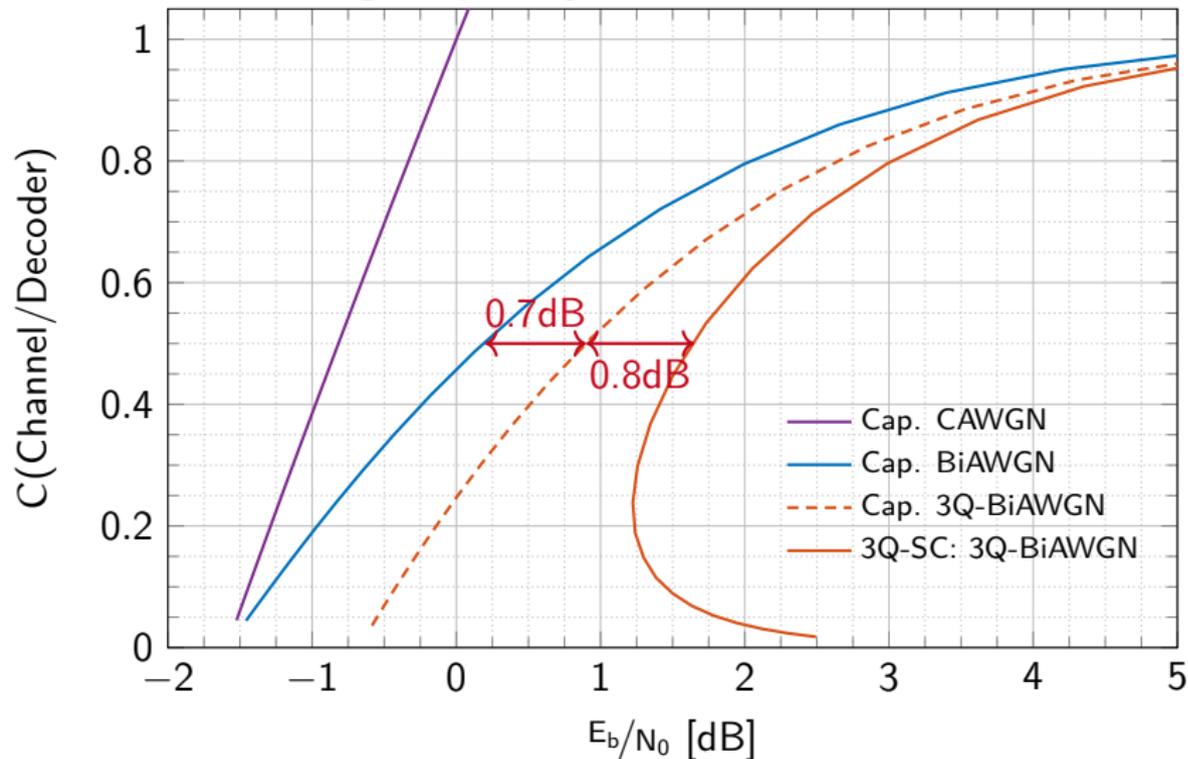
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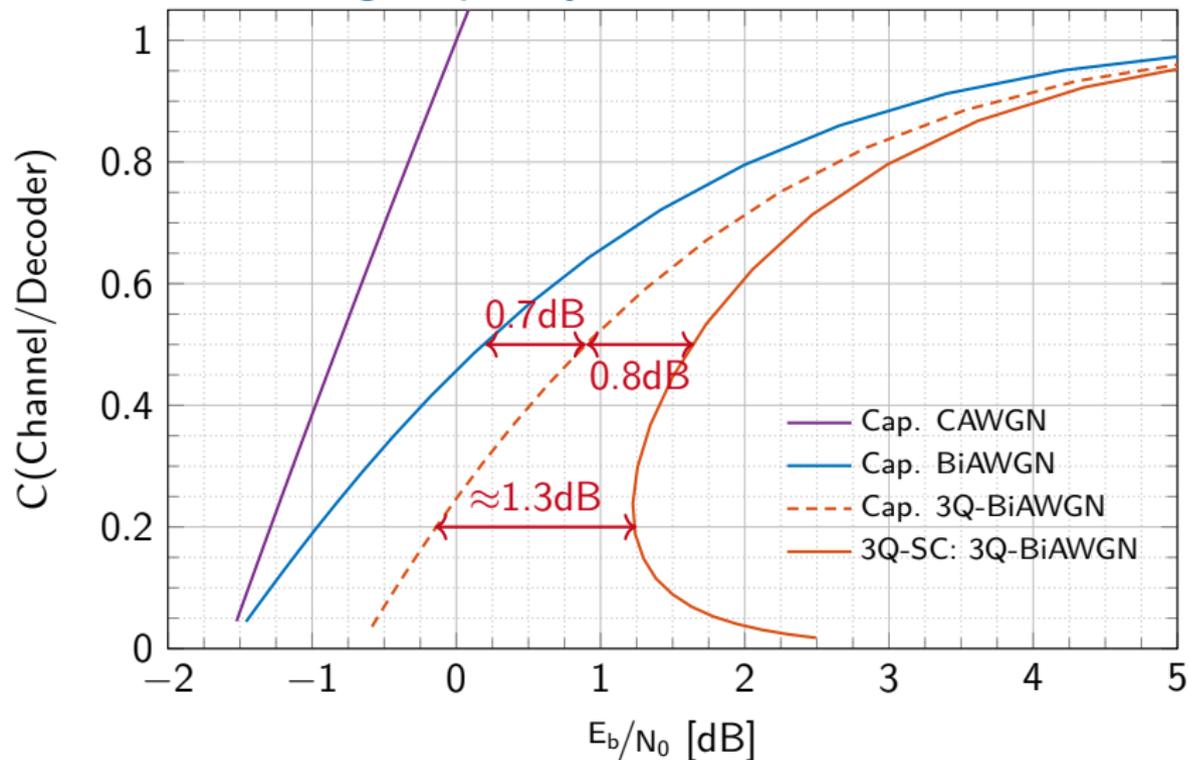
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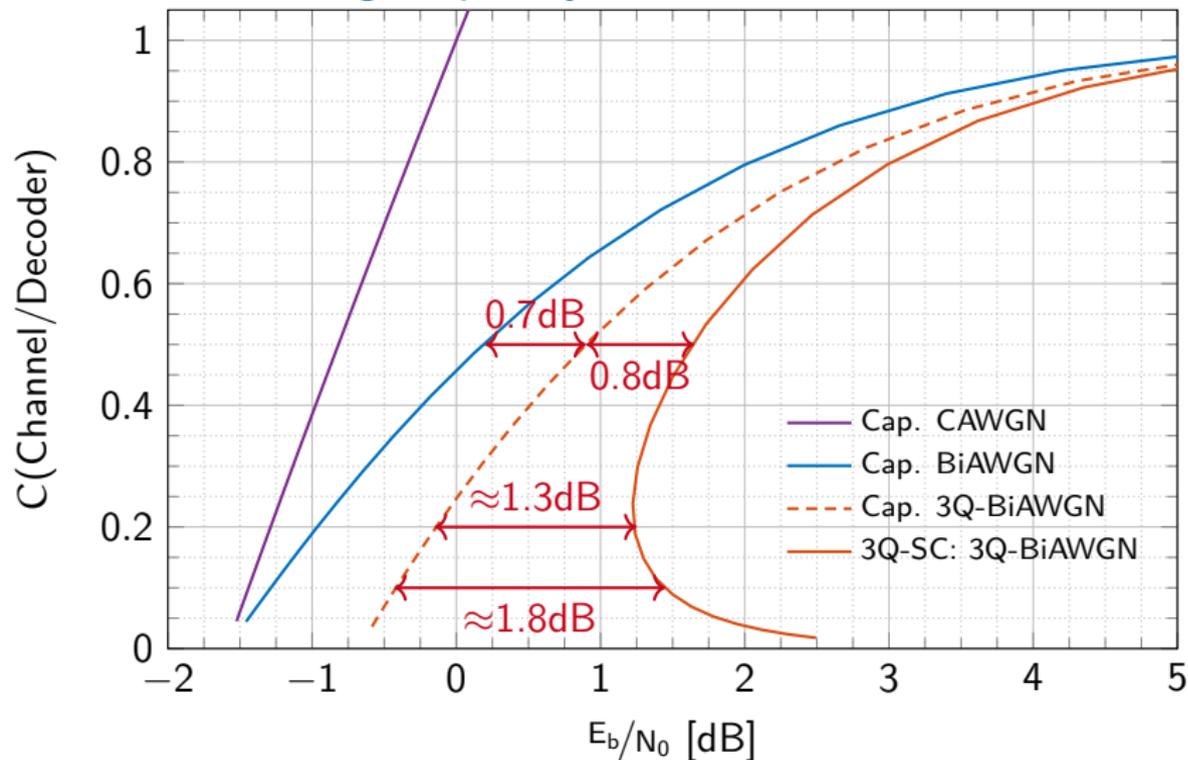
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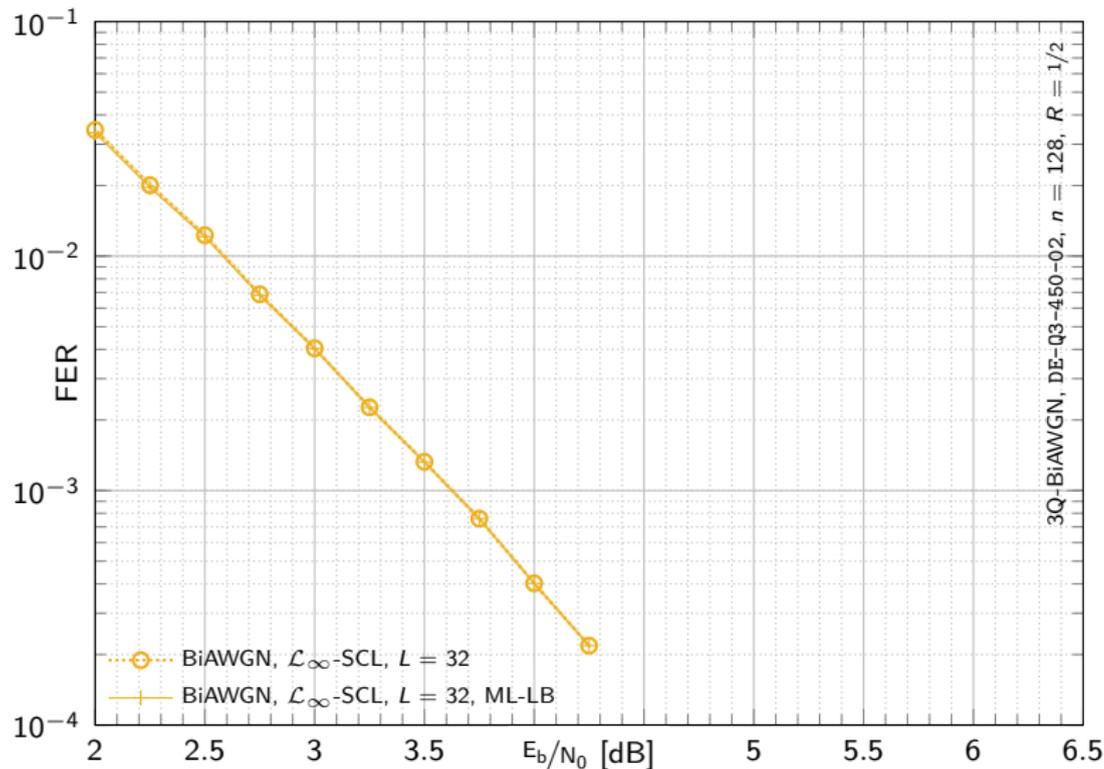
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- Use statistical reliability info in *expected path metric updates!*

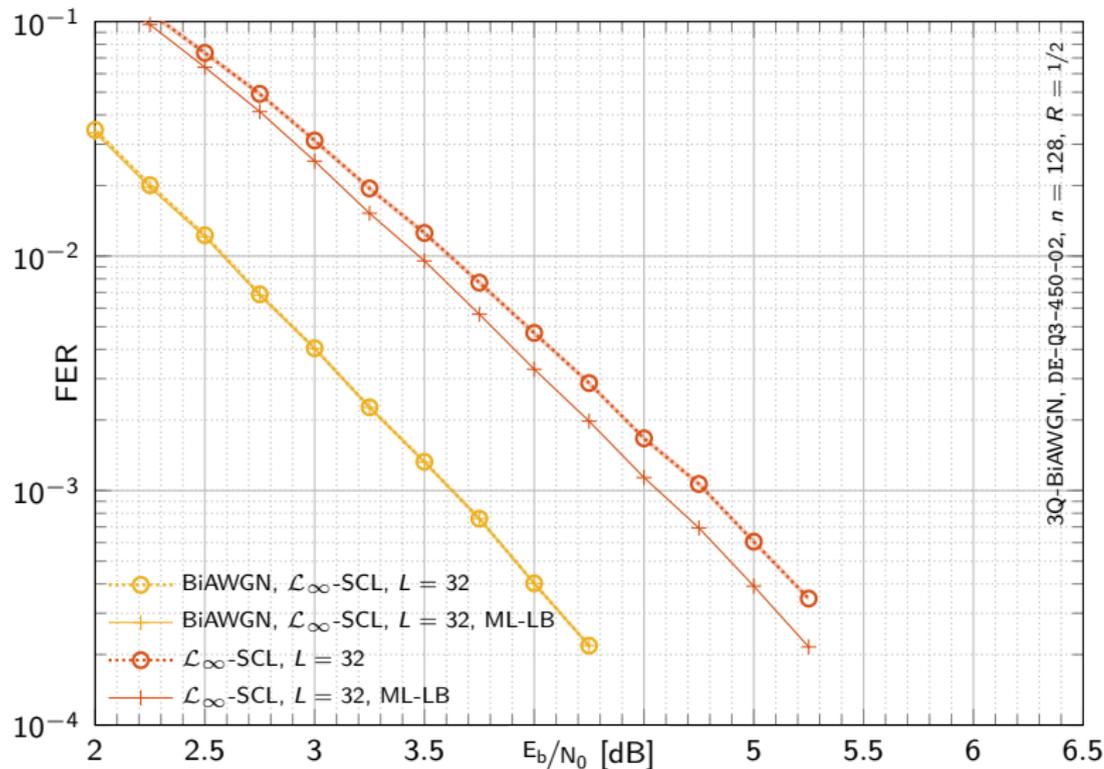


Mitigation Techniques

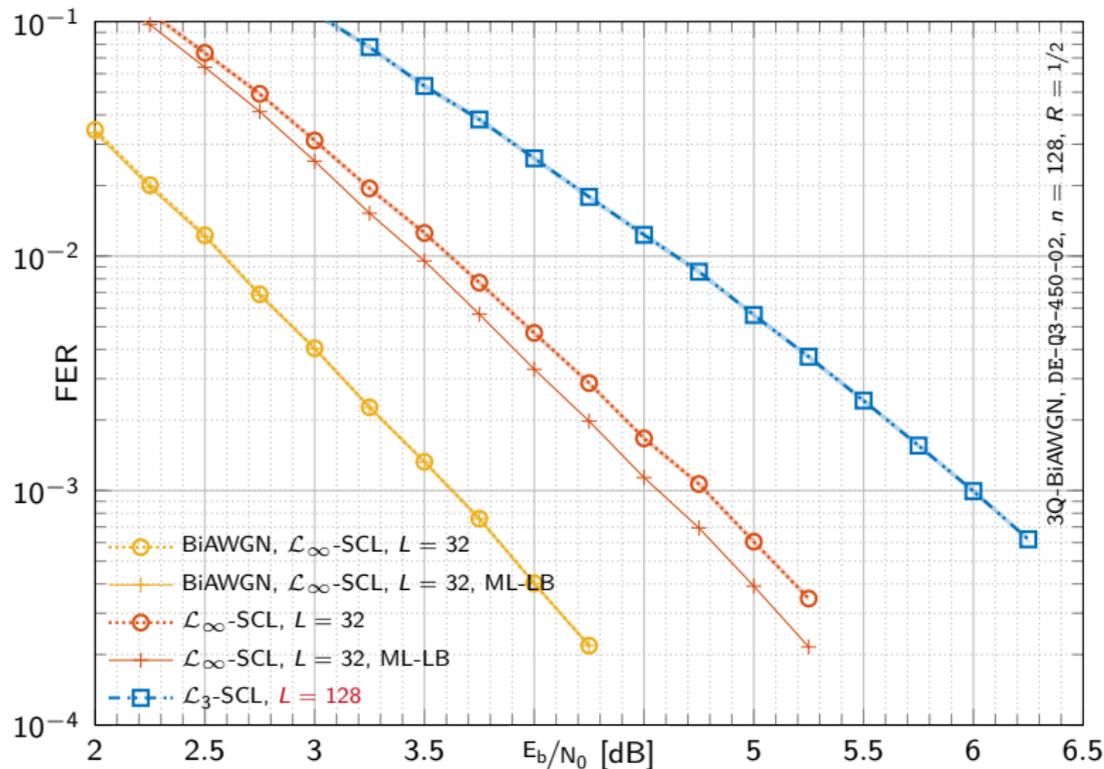
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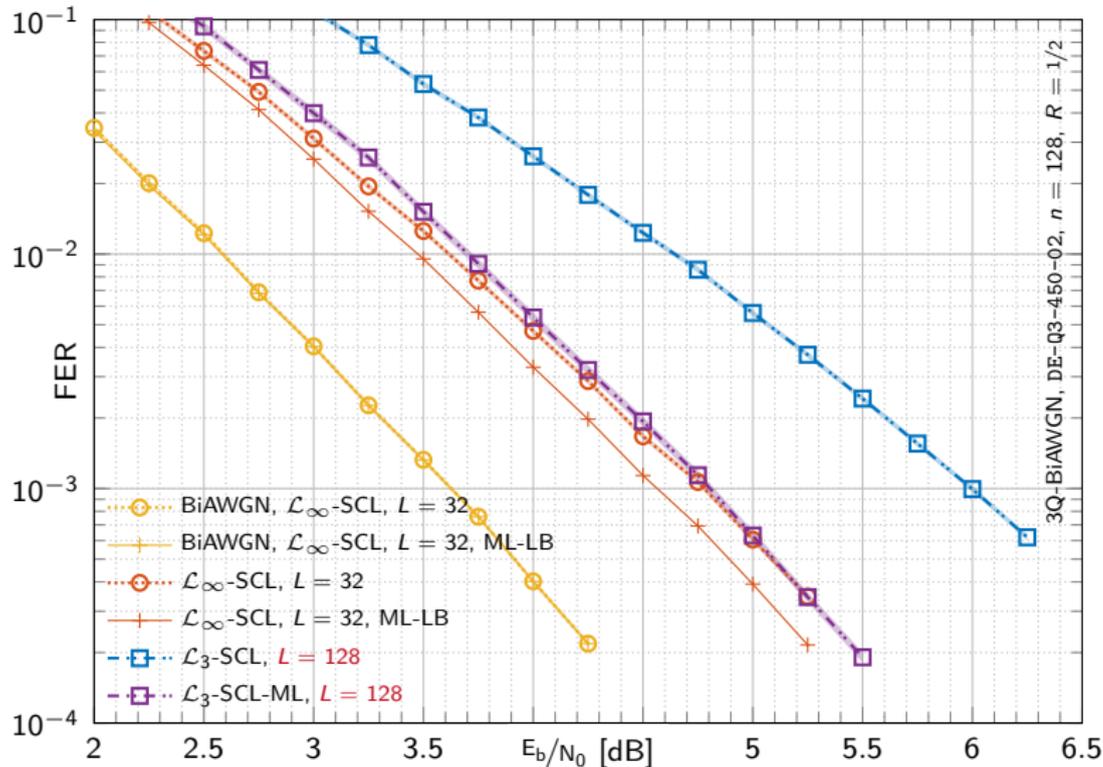
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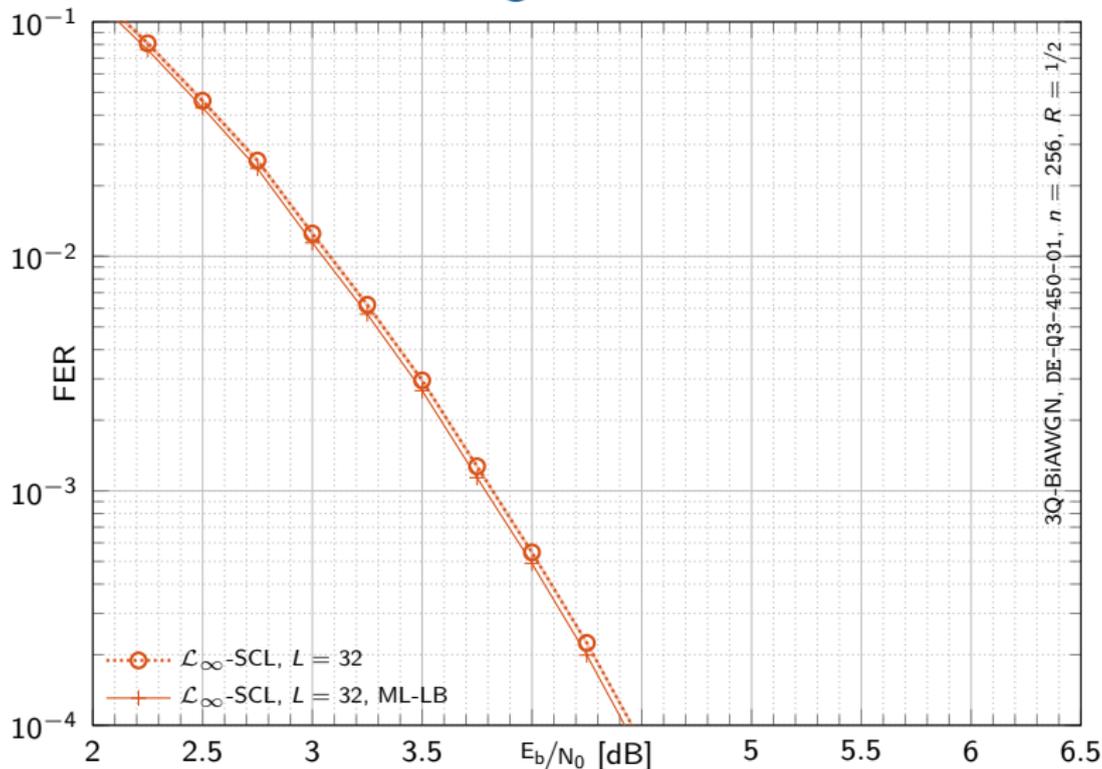
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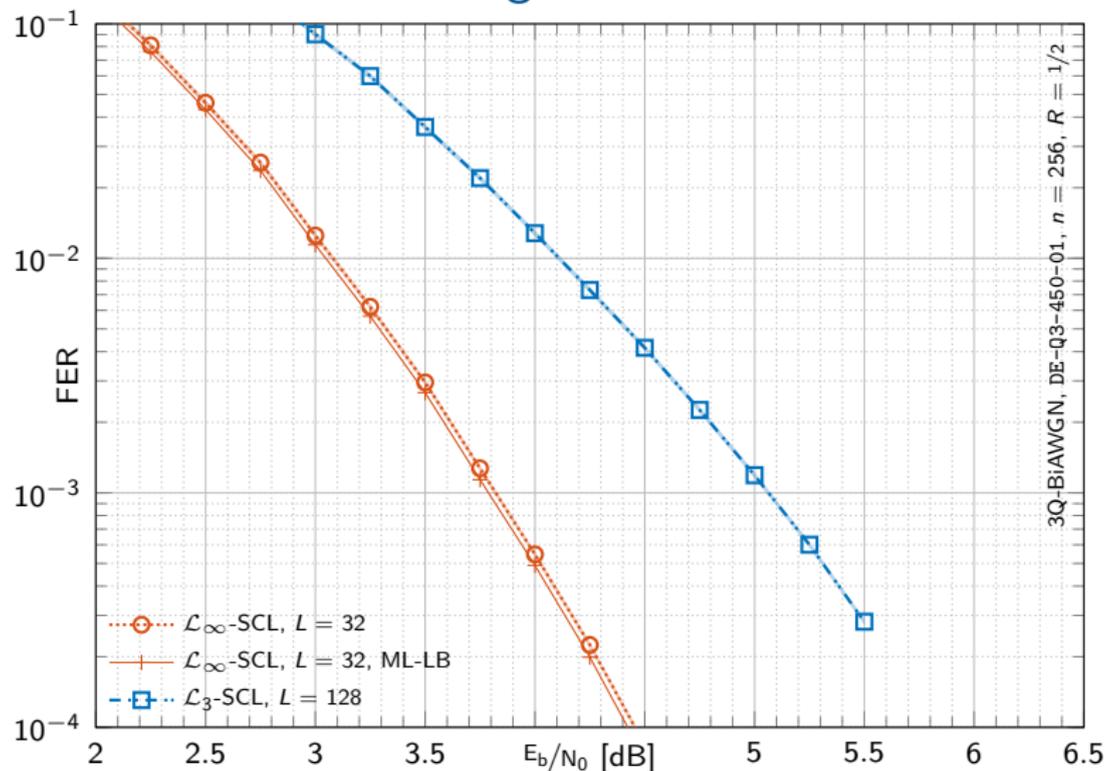
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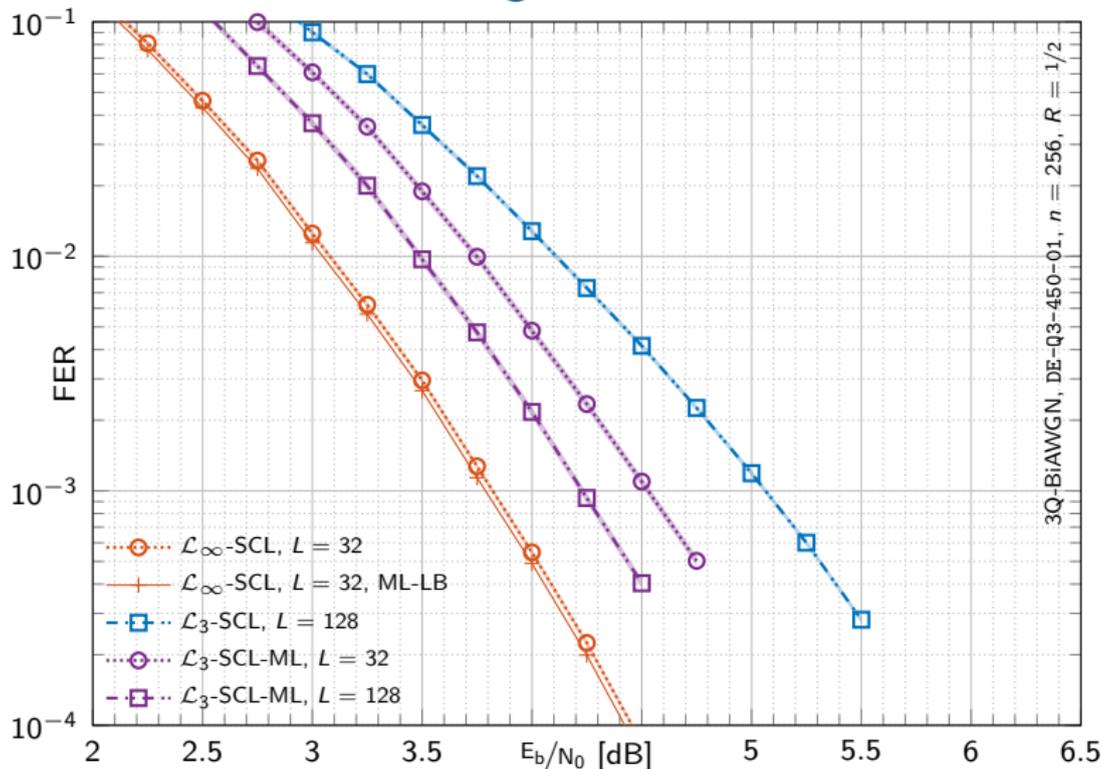
Maximum-Likelihood among List: $n = 256$



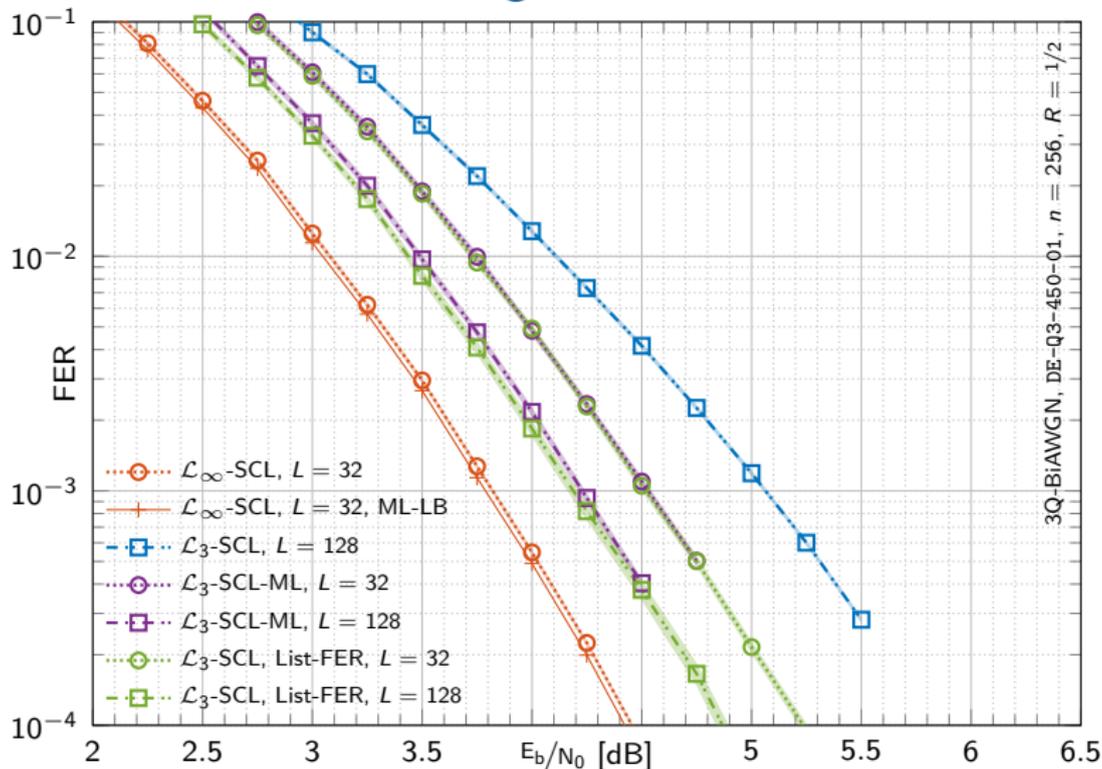
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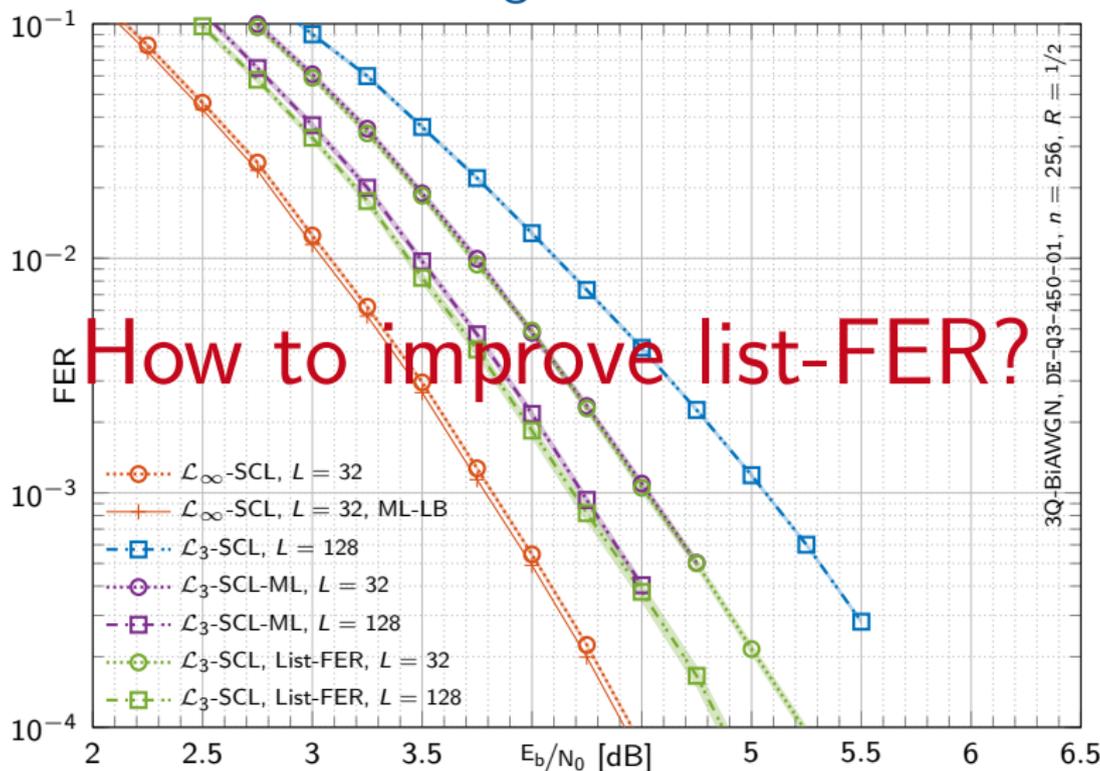
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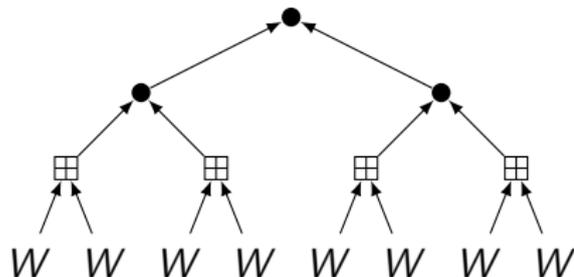




Expected Path Metric Updates: Idea

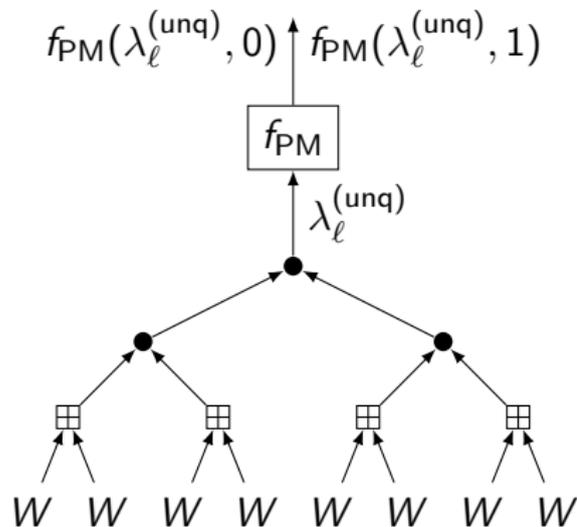
Expected Path Metric Updates: Idea

Unquantized decoder:



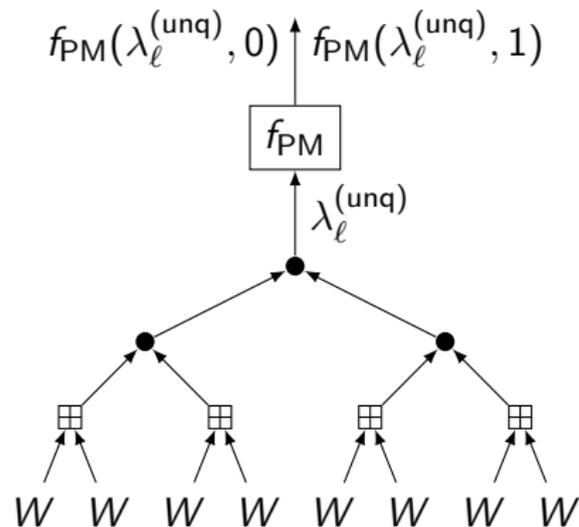
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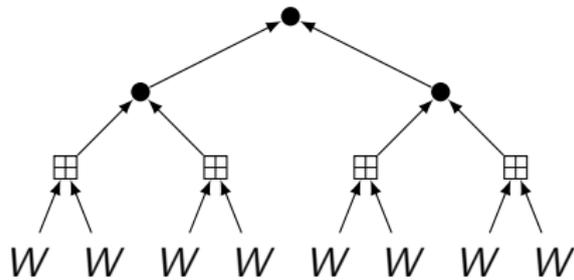


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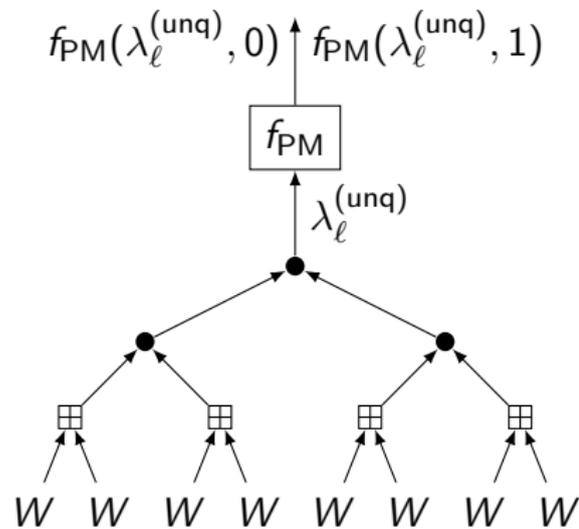


Quantized decoder:

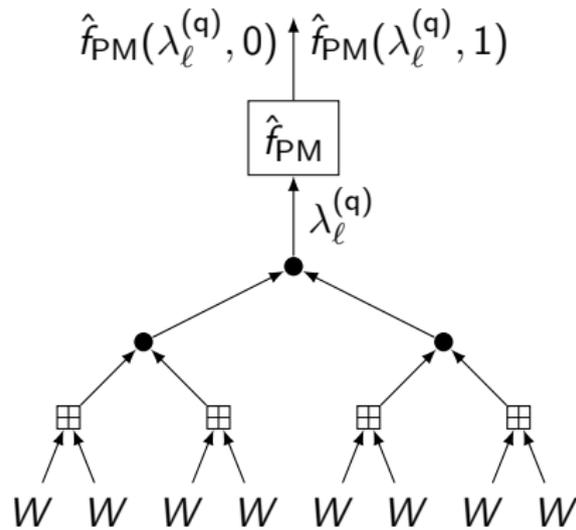


Expected Path Metric Updates: Idea

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Quantized decoder:



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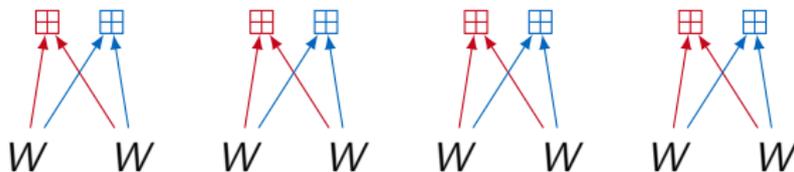
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→ How to obtain $P(\Lambda^{(\text{unq})}, \Lambda^{(\text{q})})$?

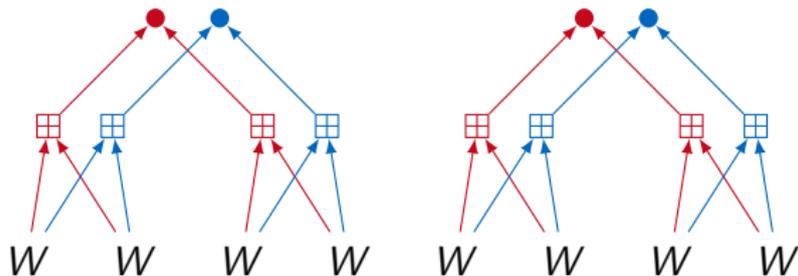
Expected Path Metric Updates: Density Evolution

W W W W W W W W

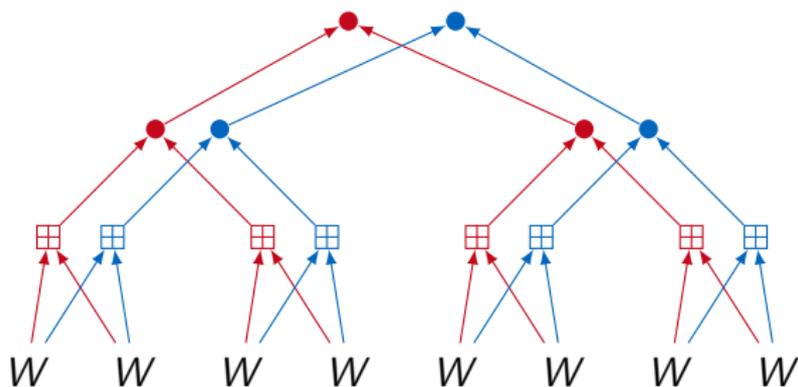
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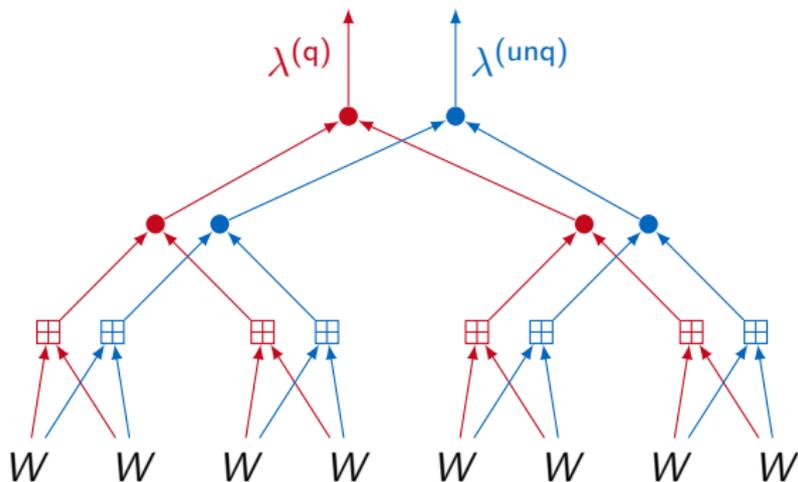
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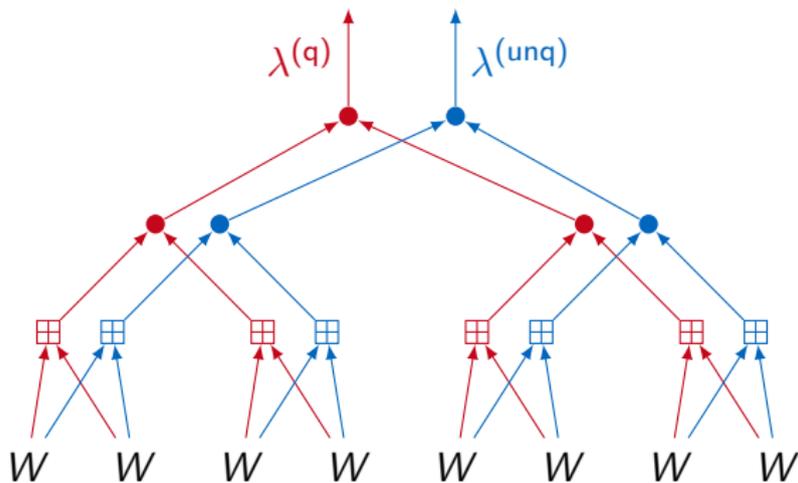
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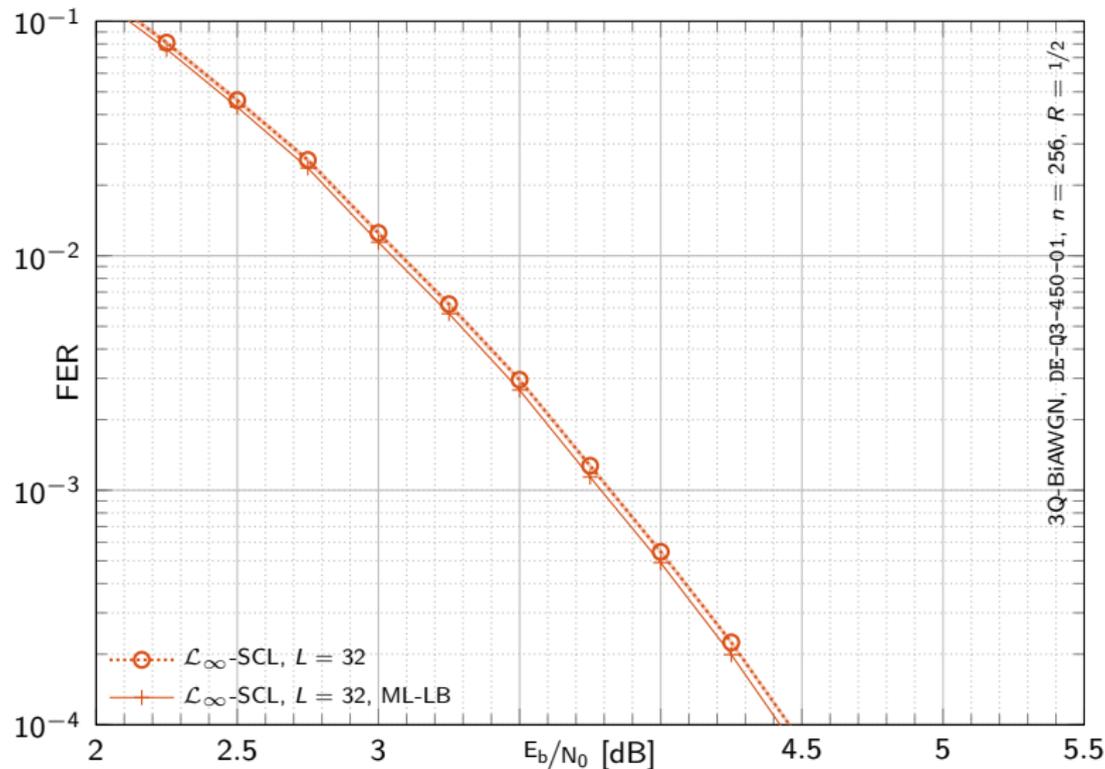


For $\lambda_1, \lambda_2 \in \mathcal{L}_{(3,\infty)} \triangleq \mathcal{L}_3 \times \mathcal{L}_\infty$:

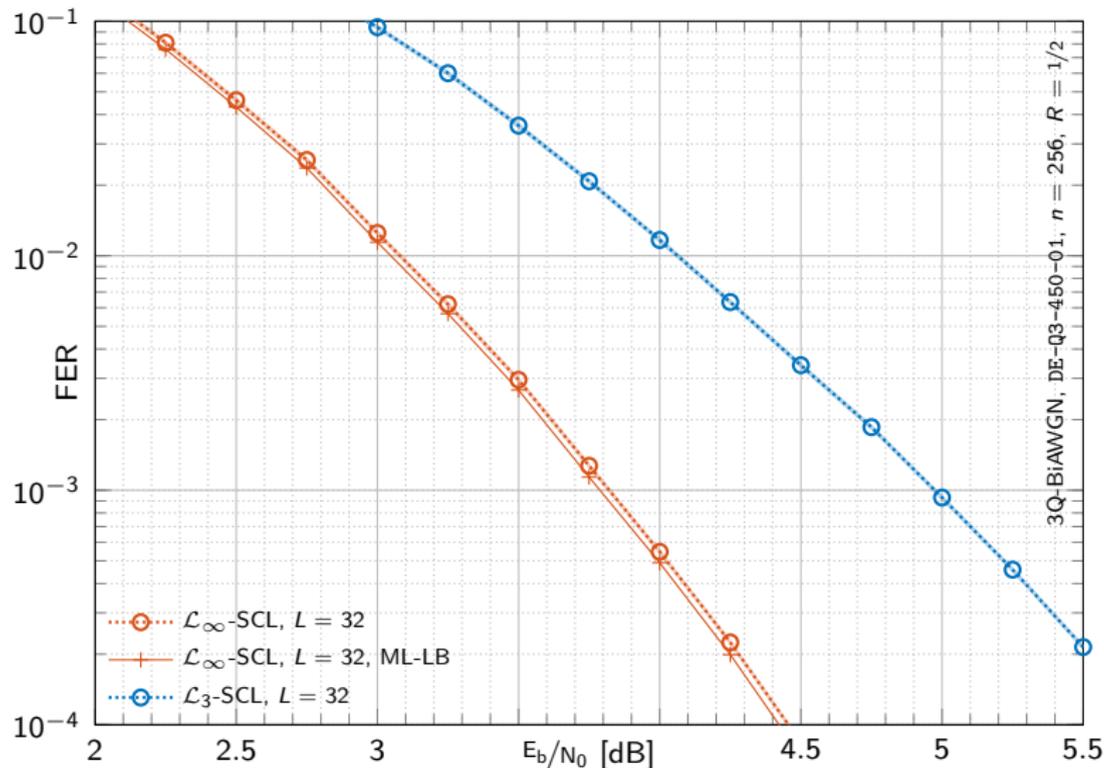
$$\lambda_1 \boxplus \lambda_2 = \left(\lambda_1^{(q)}, \lambda_1^{(\text{unq})} \right) \boxplus \left(\lambda_2^{(q)}, \lambda_2^{(\text{unq})} \right) \triangleq \left(\lambda_1^{(q)} \boxplus \lambda_2^{(q)}, \lambda_1^{(\text{unq})} \boxplus \lambda_2^{(\text{unq})} \right)$$

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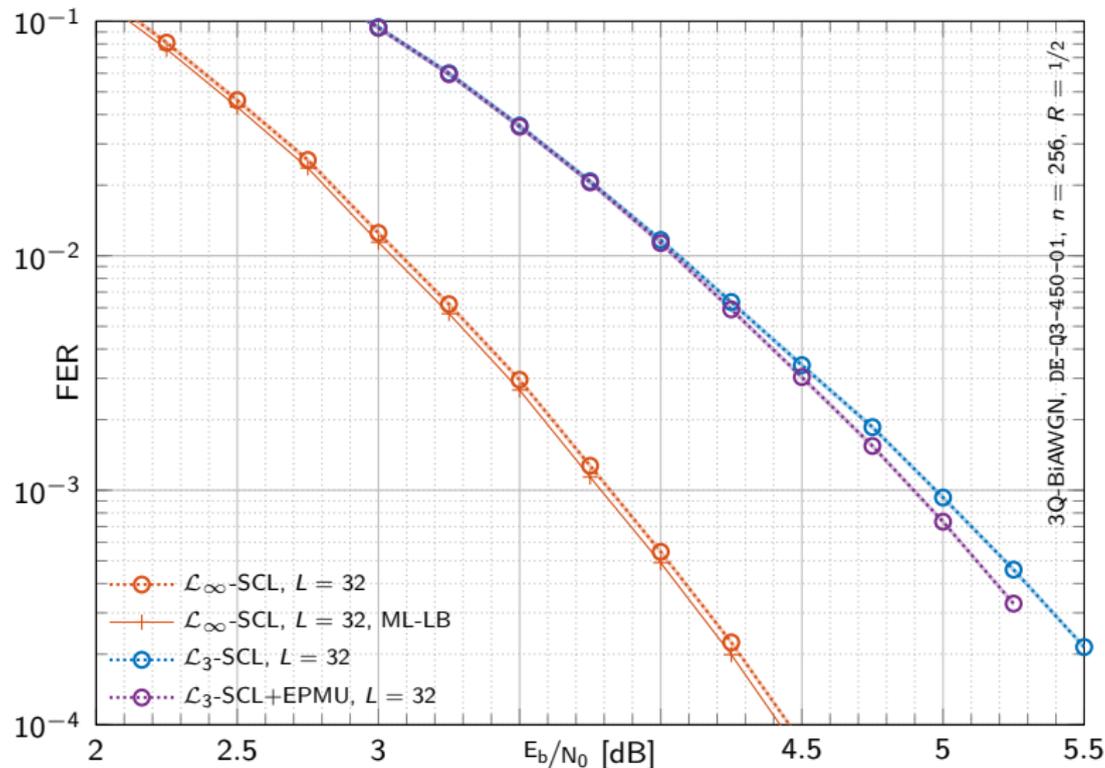
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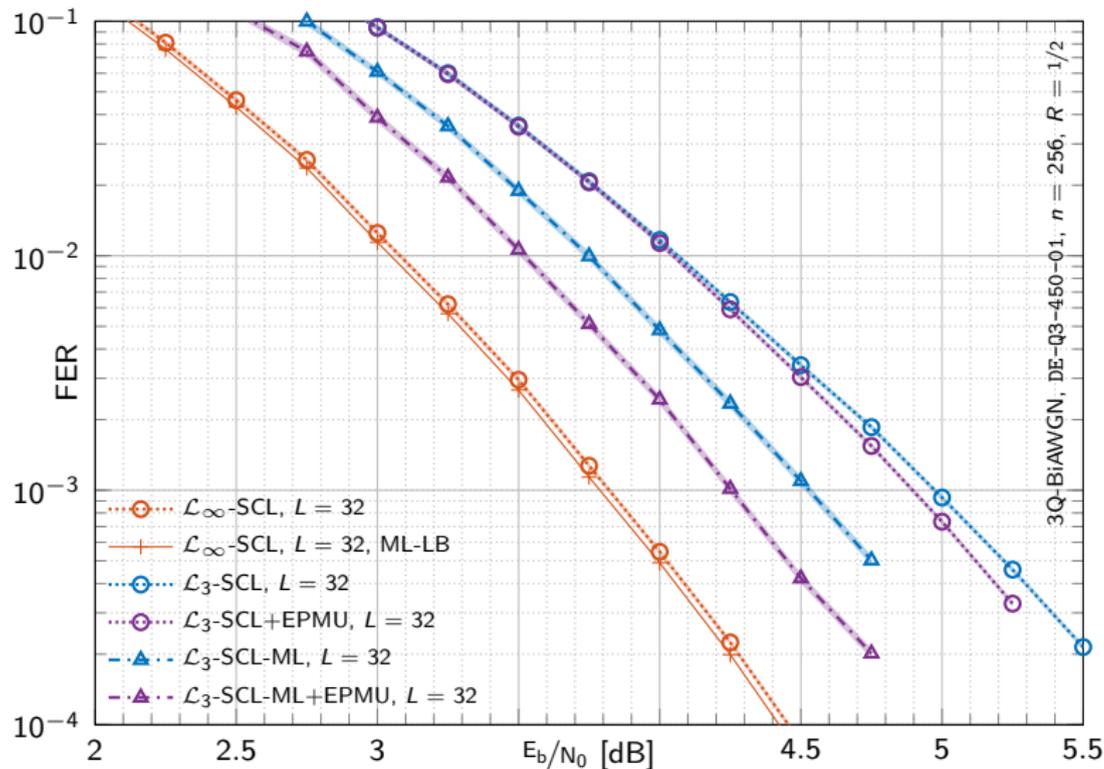
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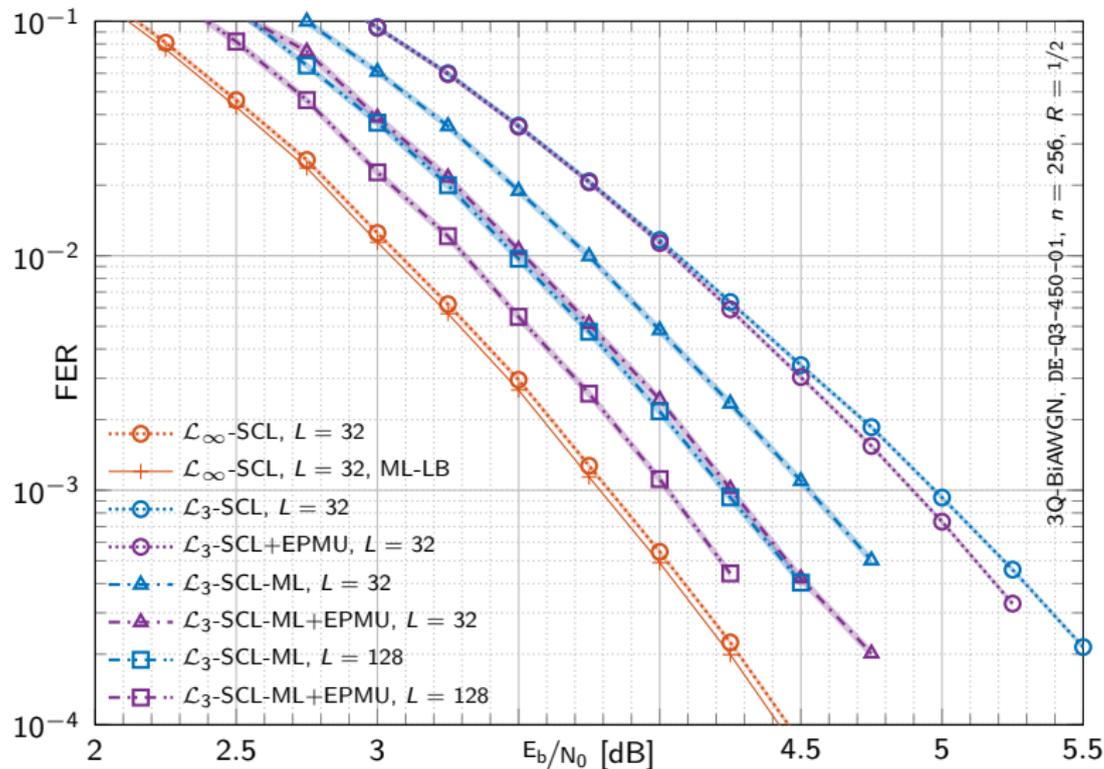
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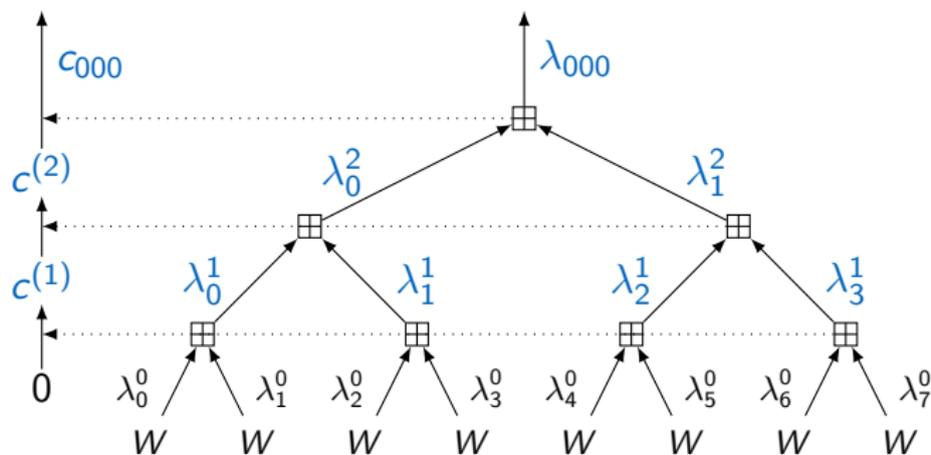


EPMU with Contradiction Counting

Idea: Count contradictions at variable nodes as low-complexity instantaneous reliability indicator!

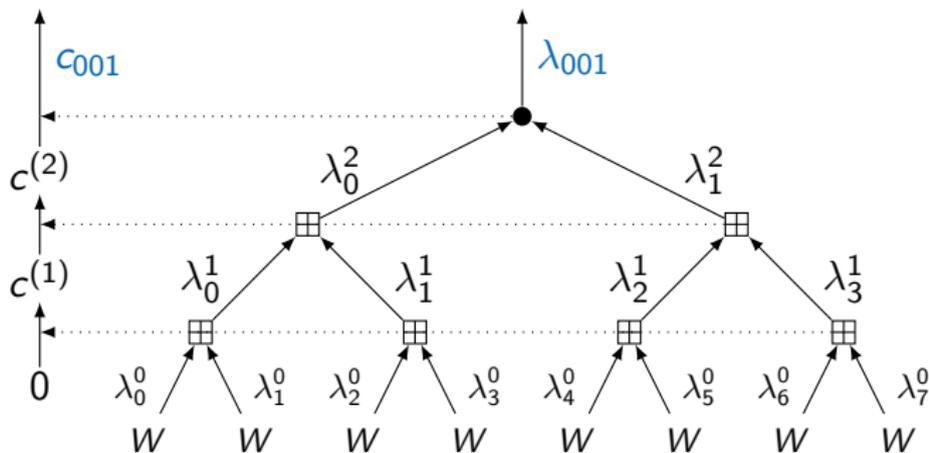
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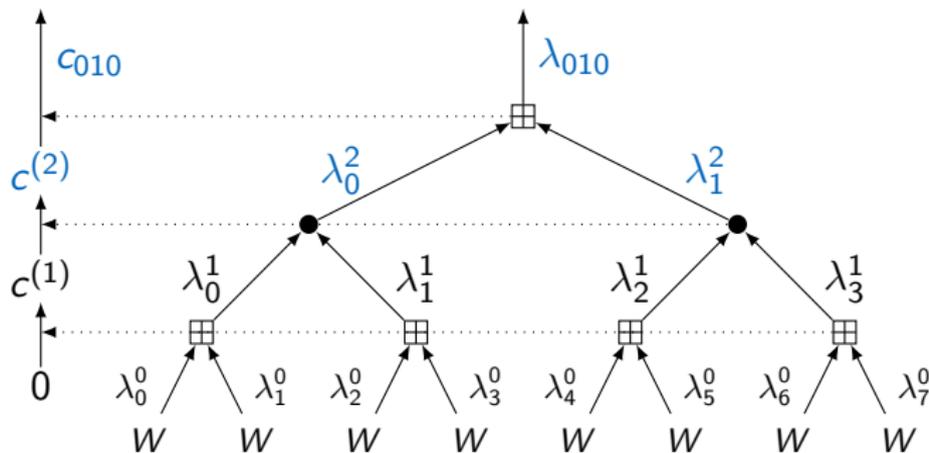
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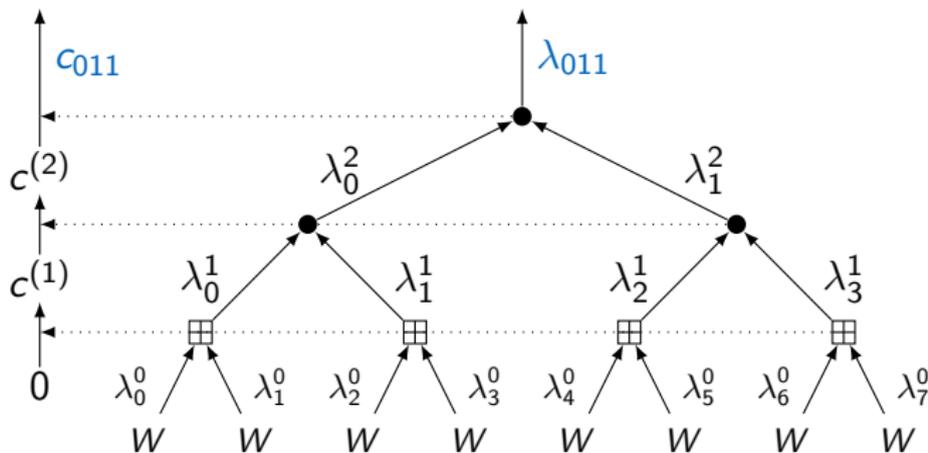
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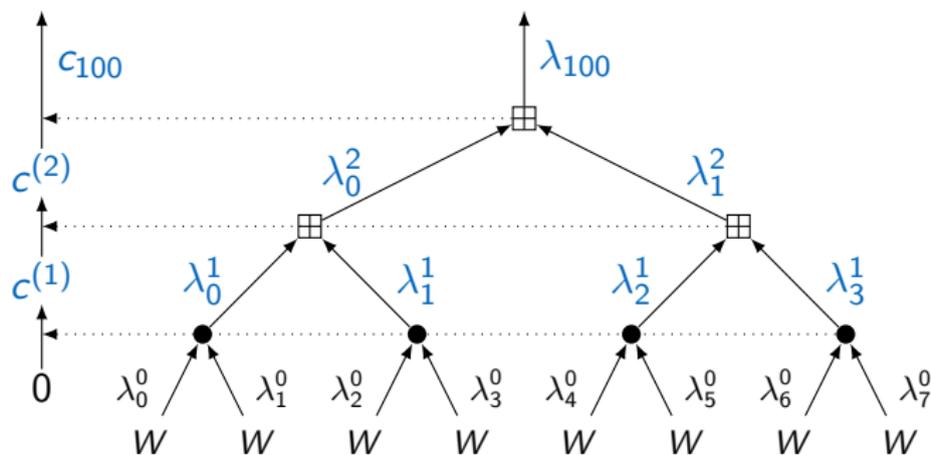
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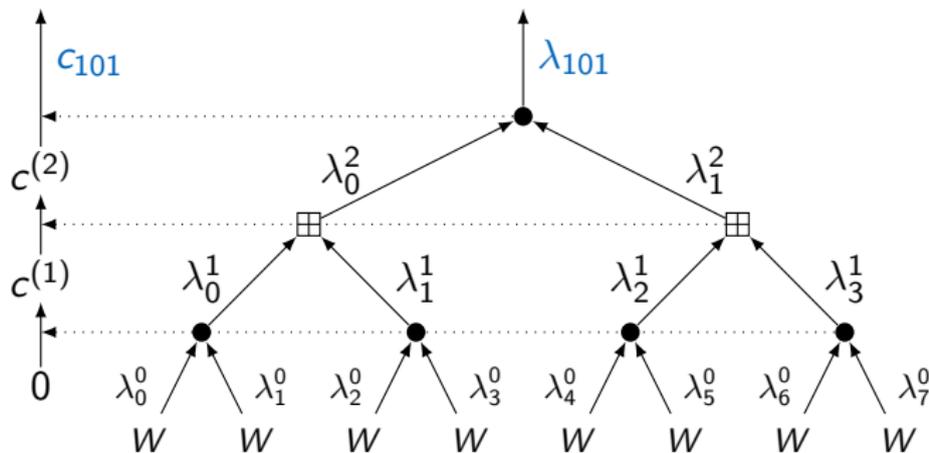
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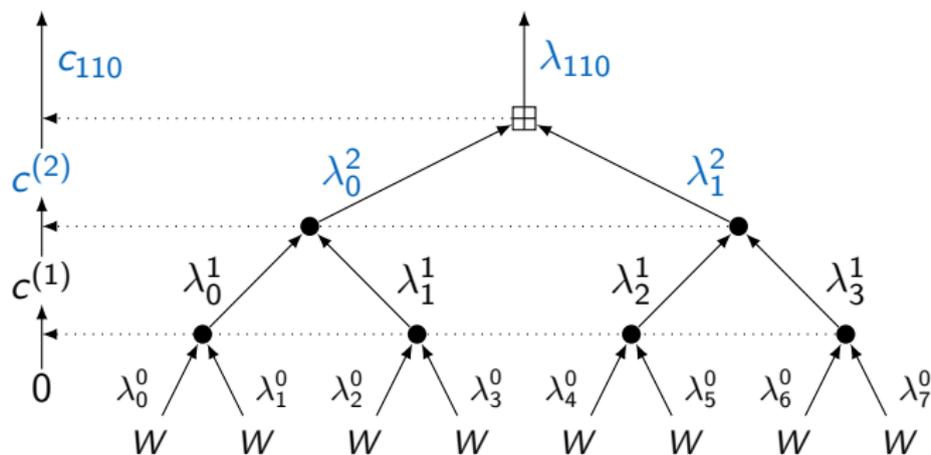
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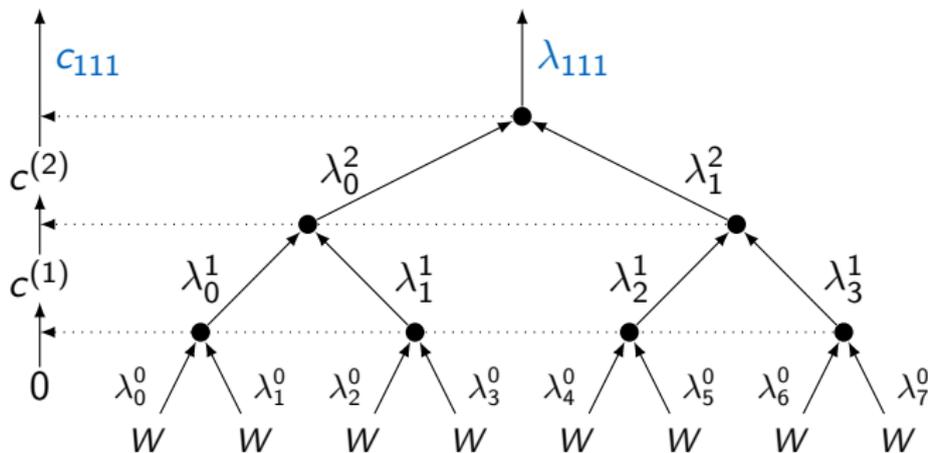
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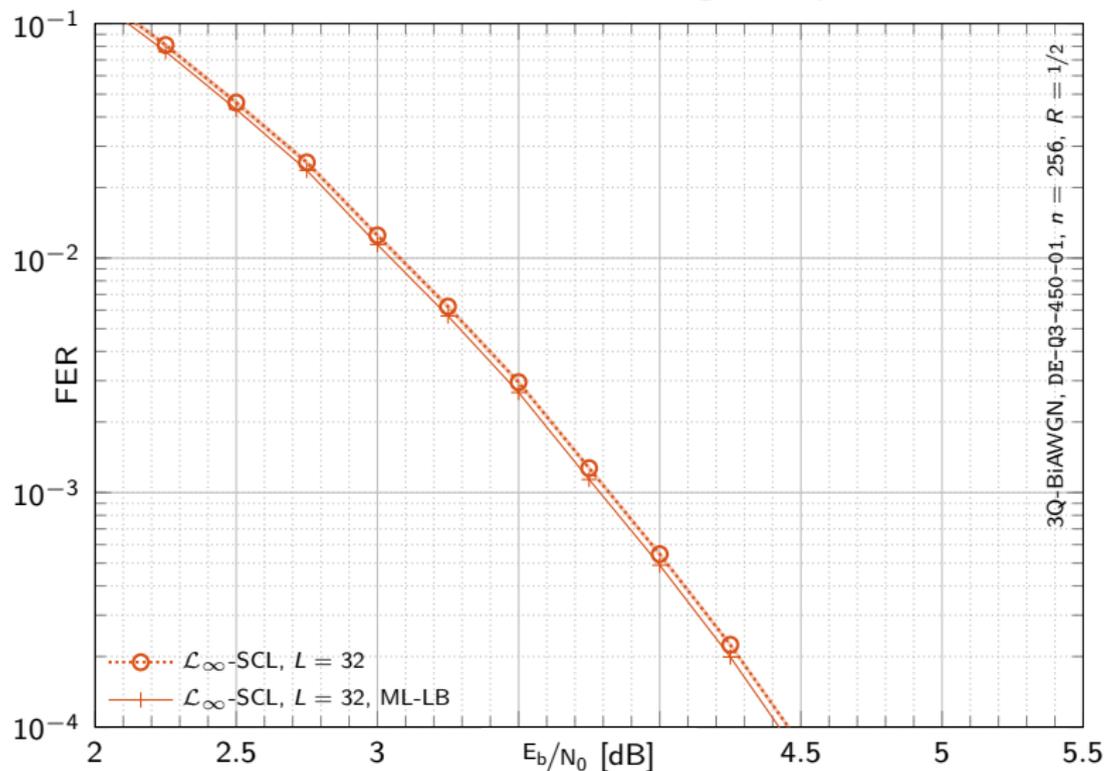


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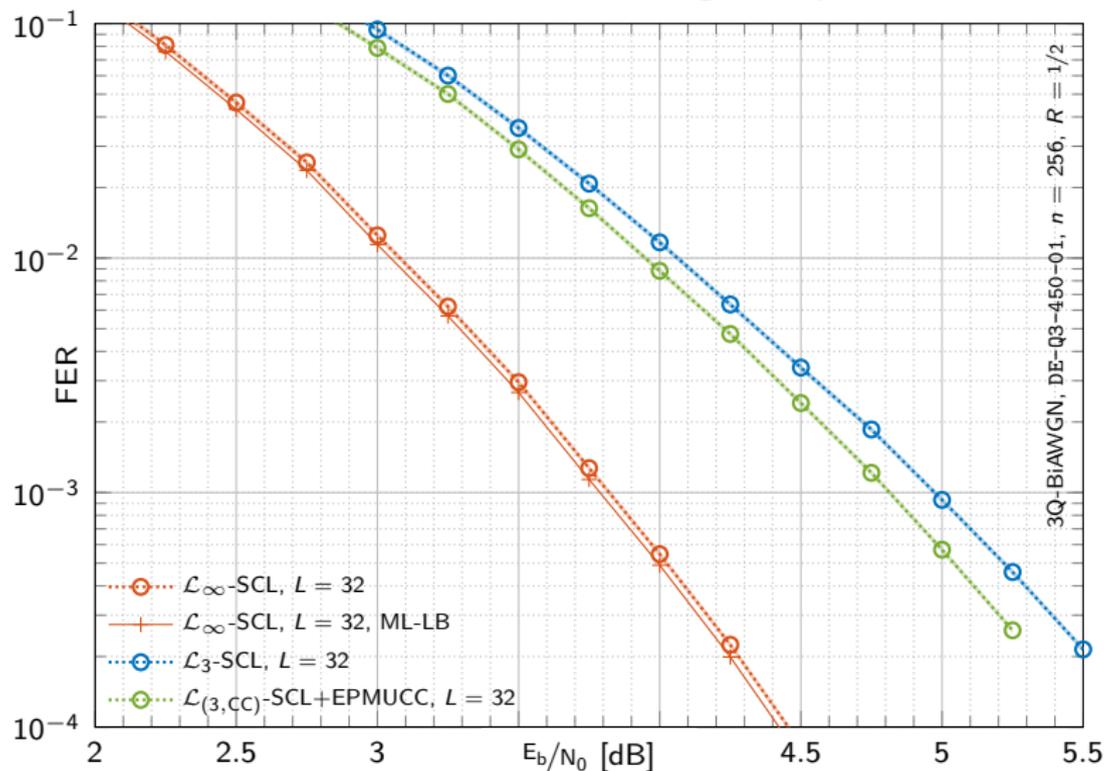
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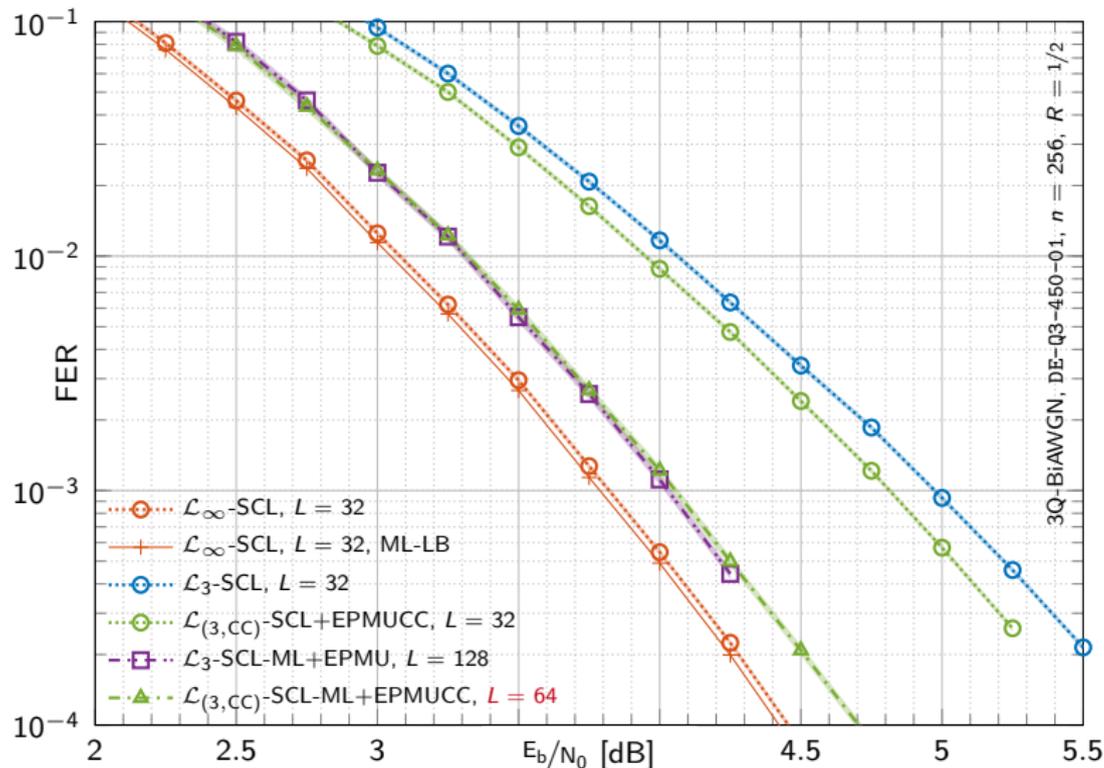
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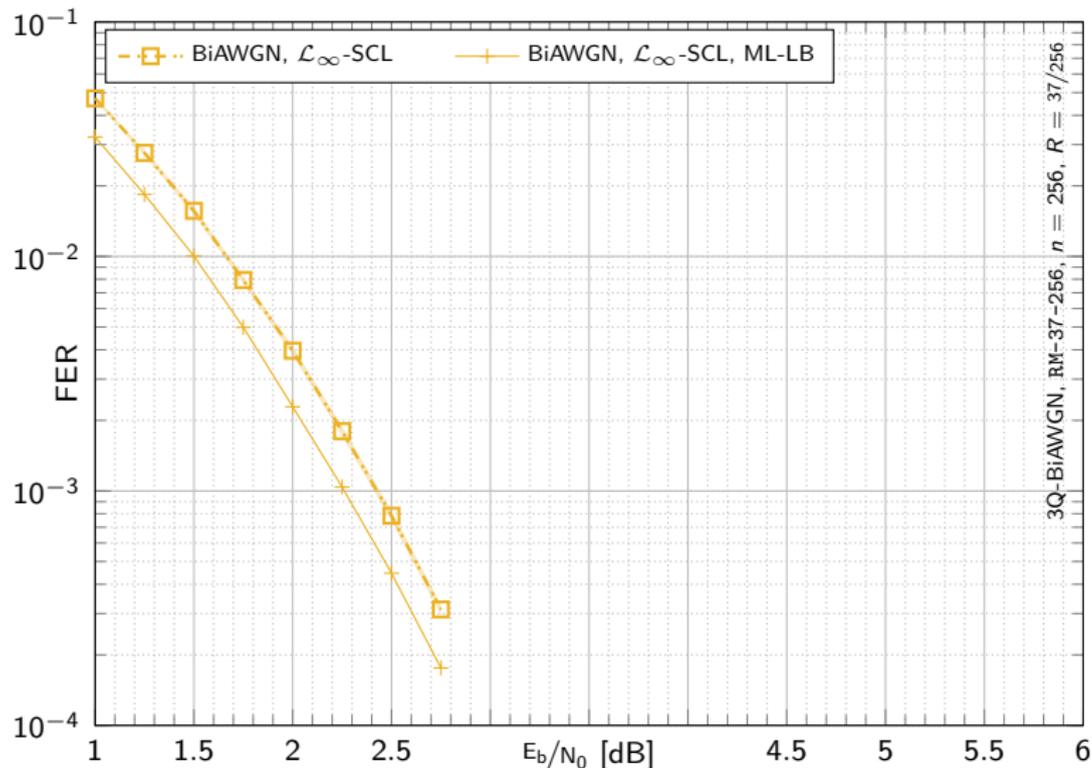
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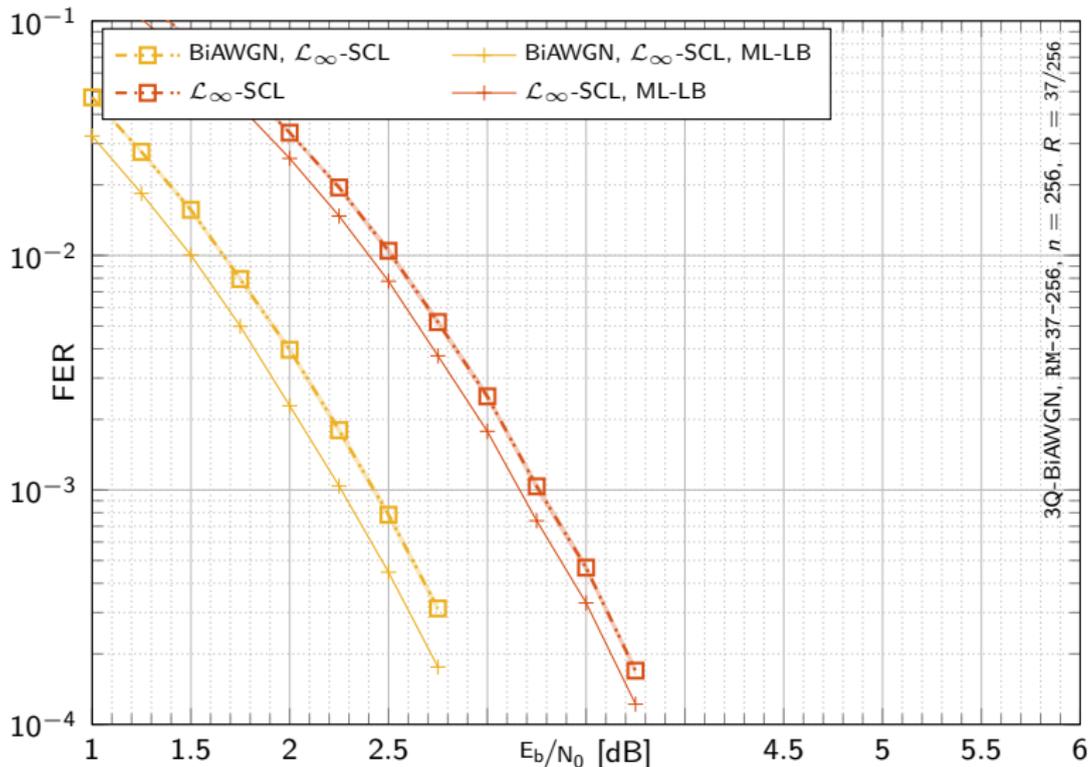


Validation / Robustness

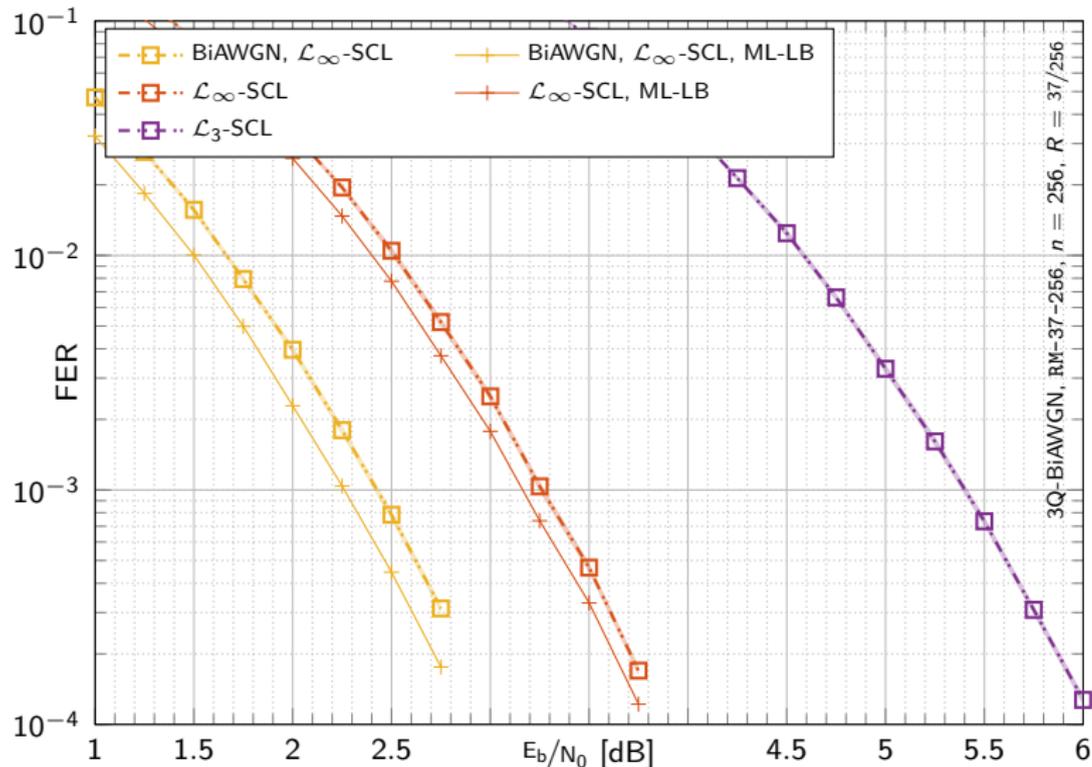
Validation: Low Code Rate: $R \approx 0.145$, $L = 128$



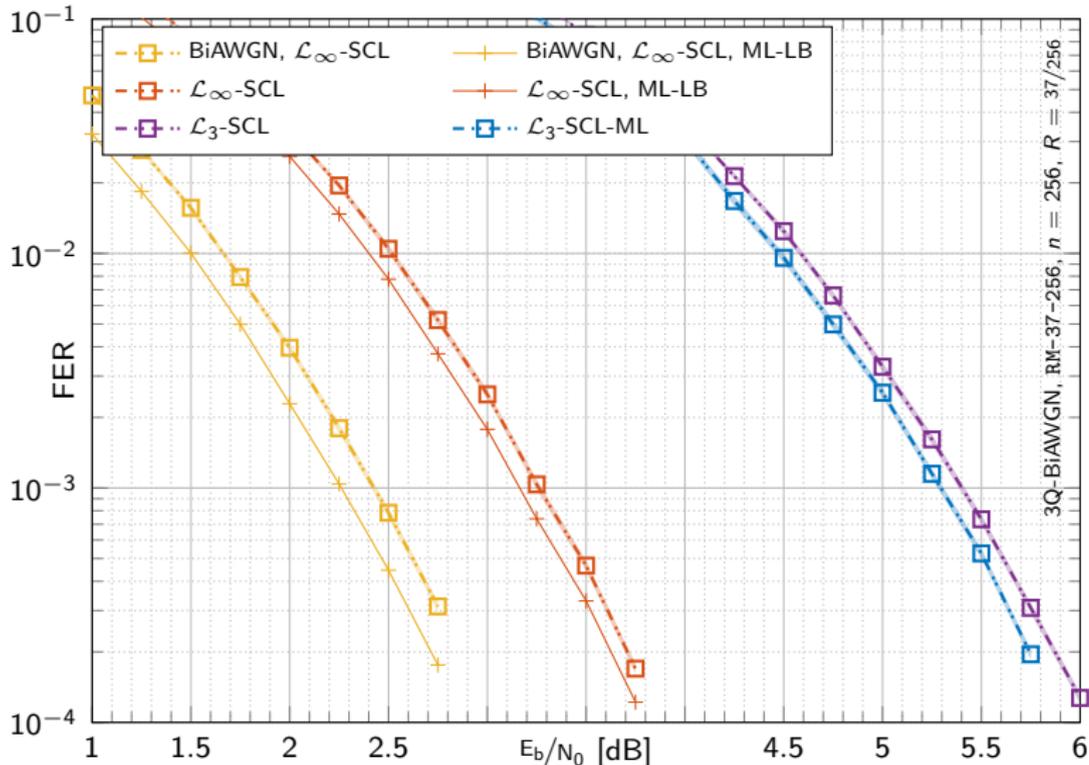
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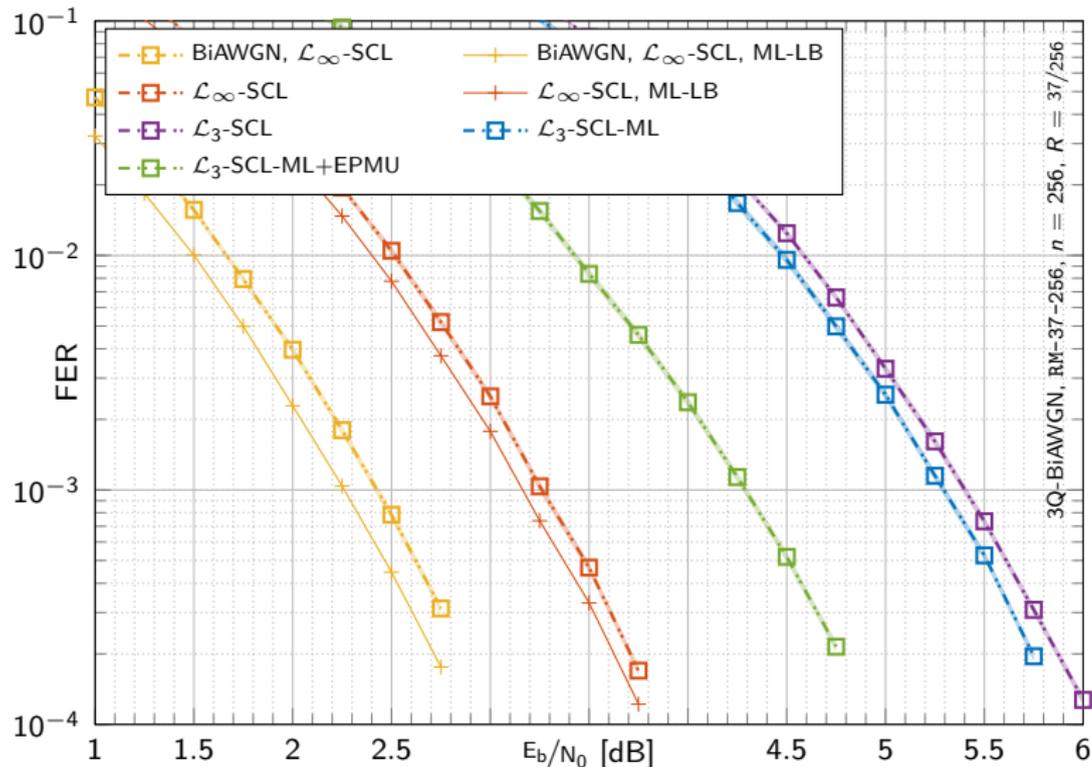
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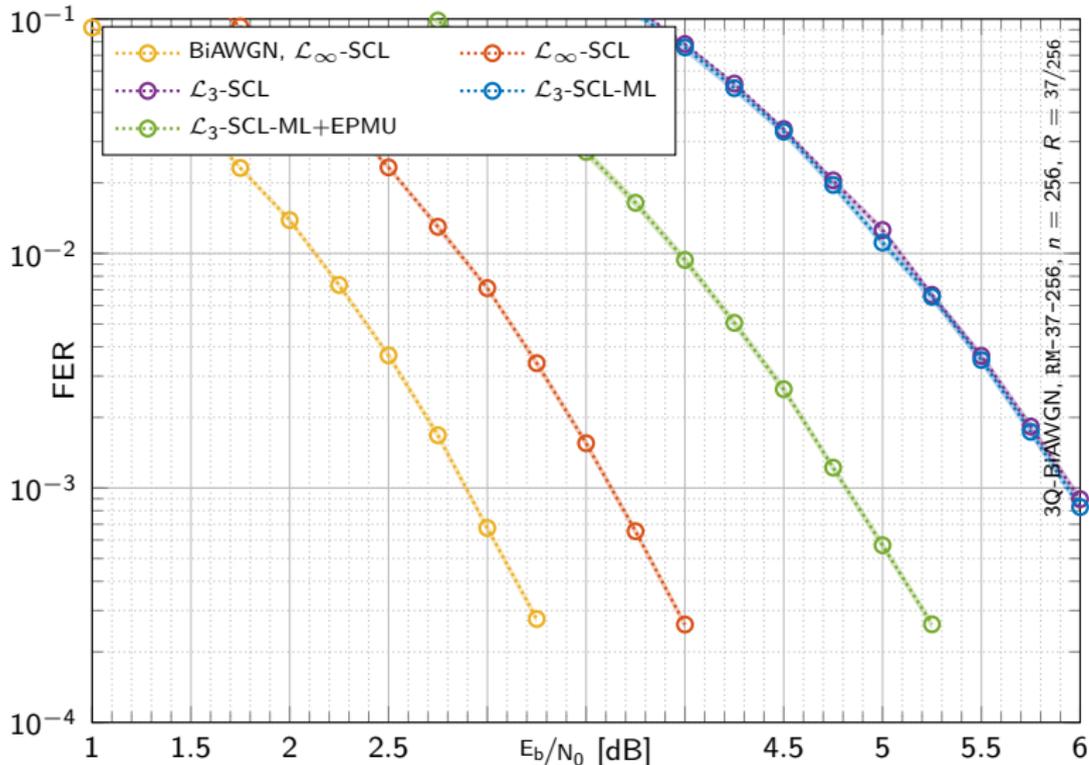
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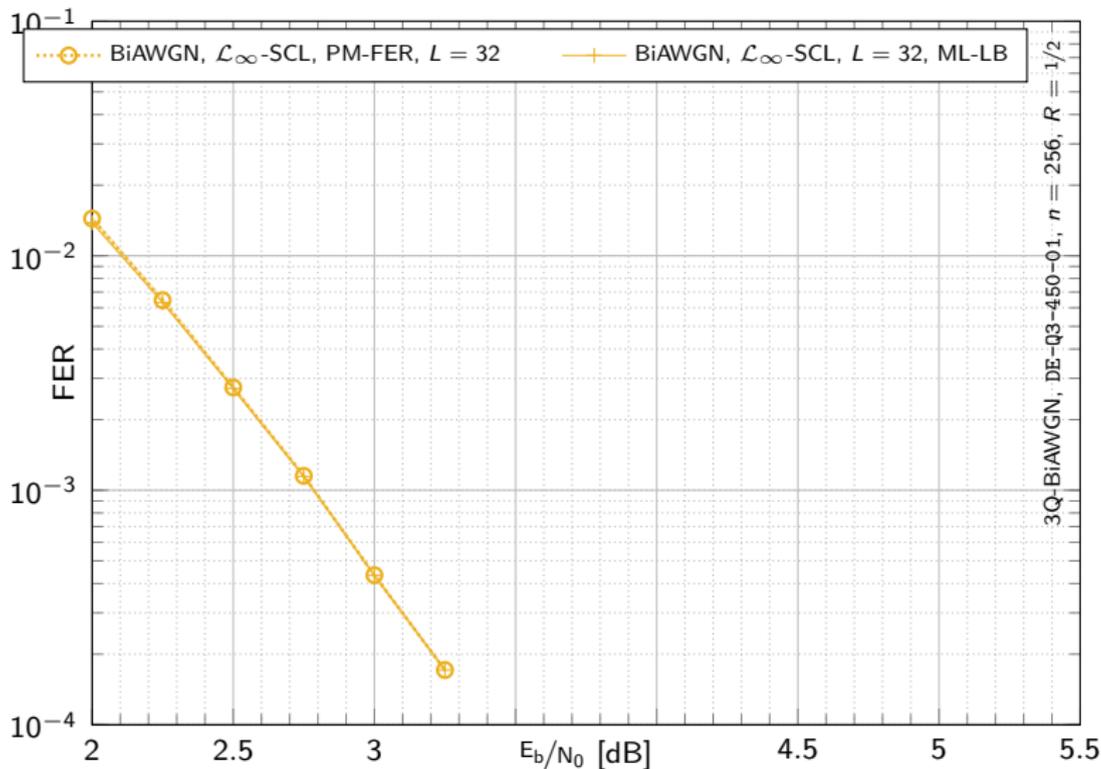
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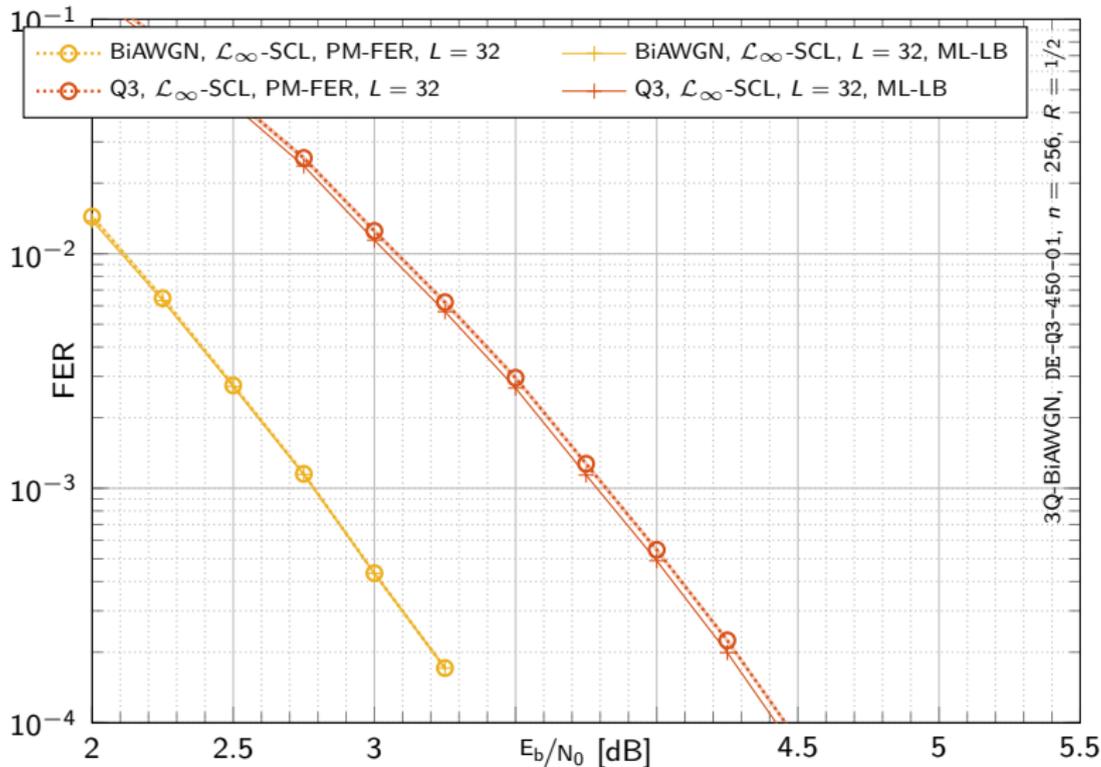
Validation: Low Code Rate: $R \approx 0.145$, $L = 32$



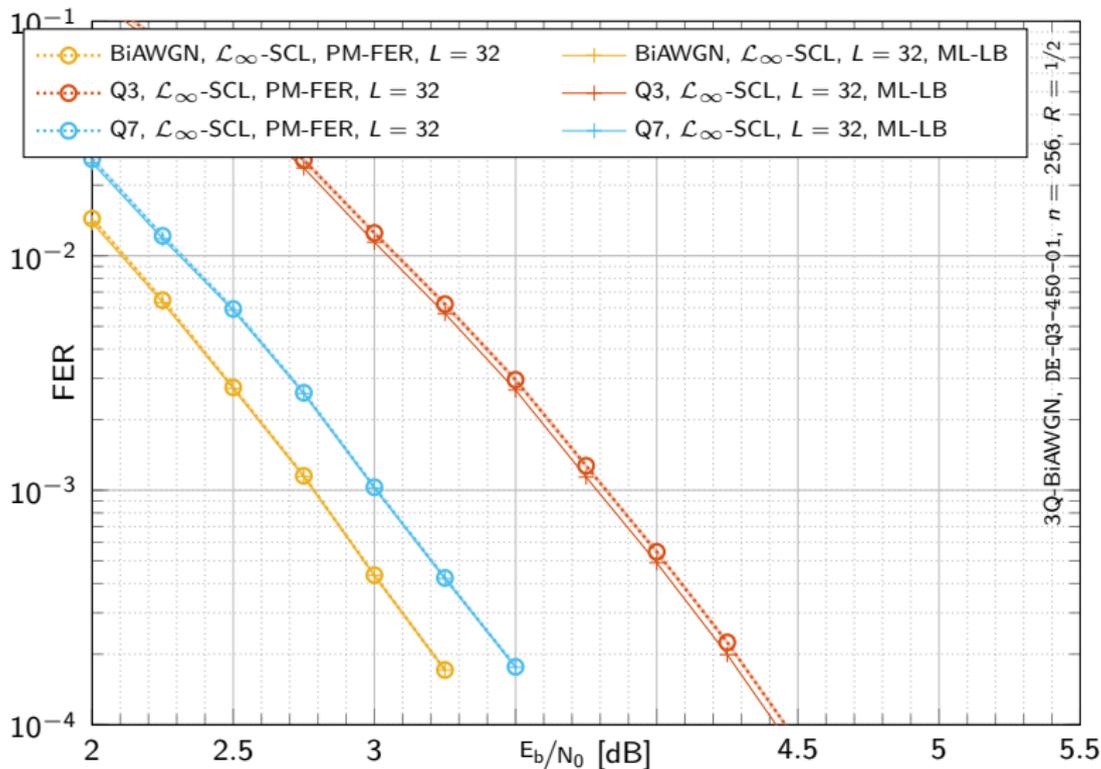
Comparison: Q7 Performance



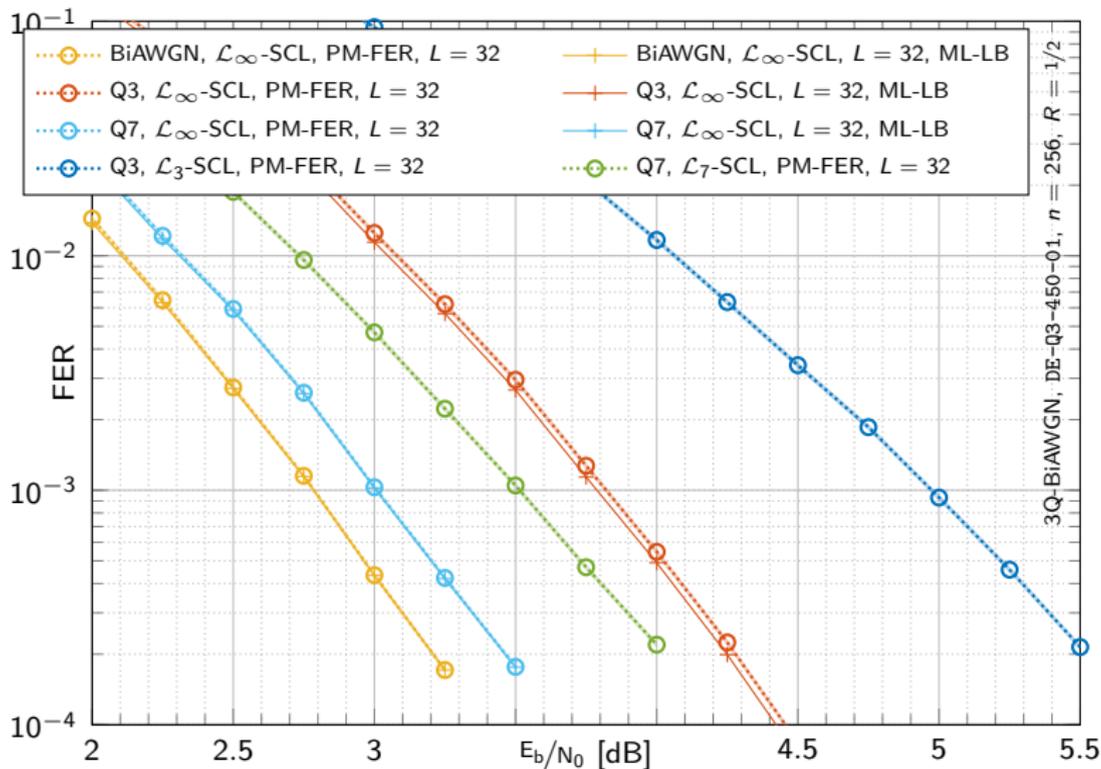
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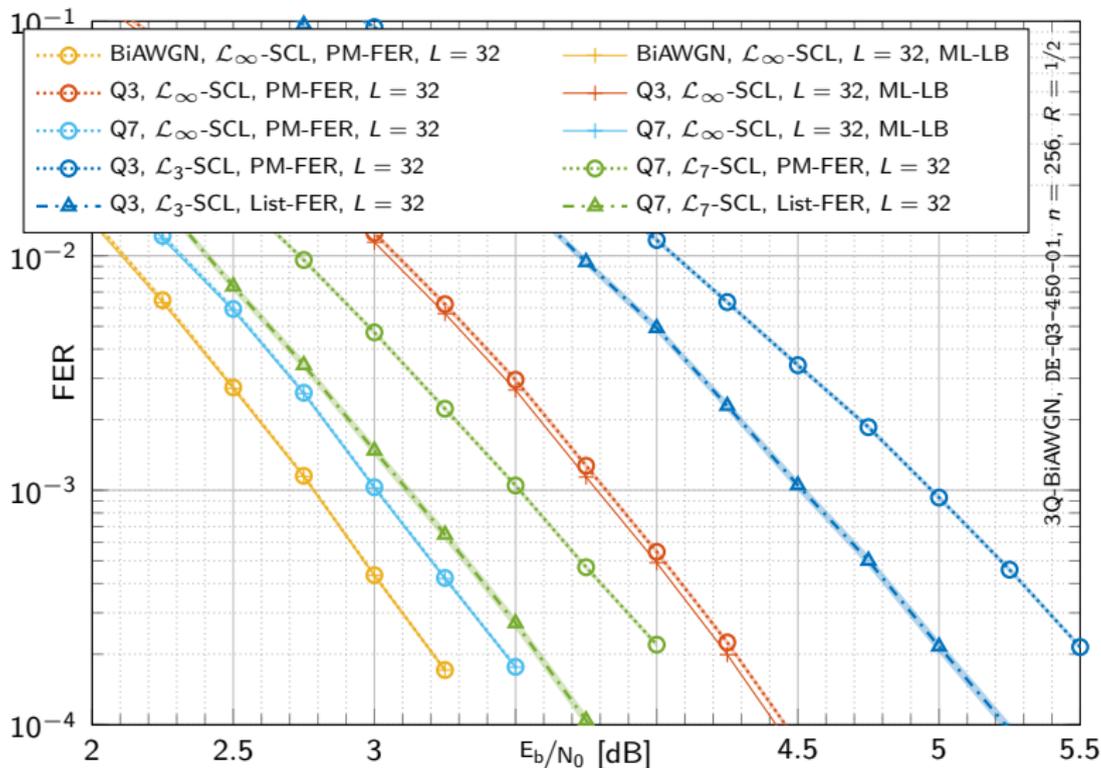
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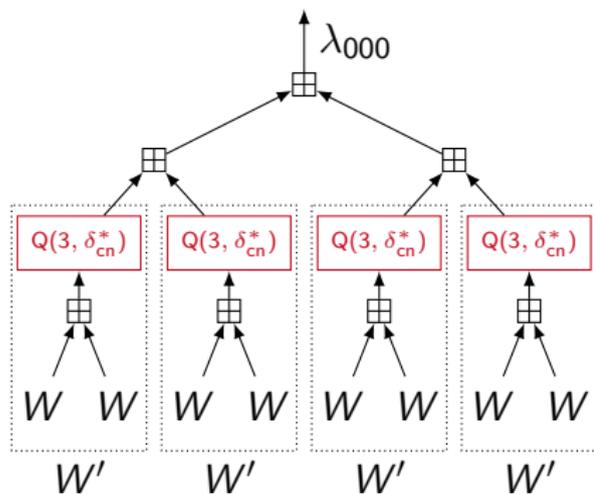




Comparison: '1st Layer Unquantized Then Q3'

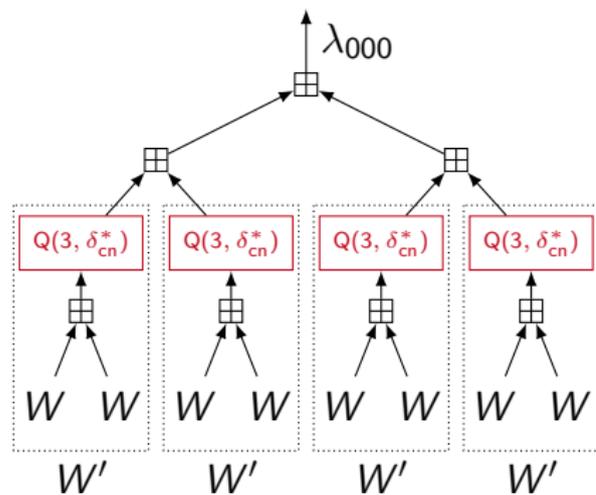
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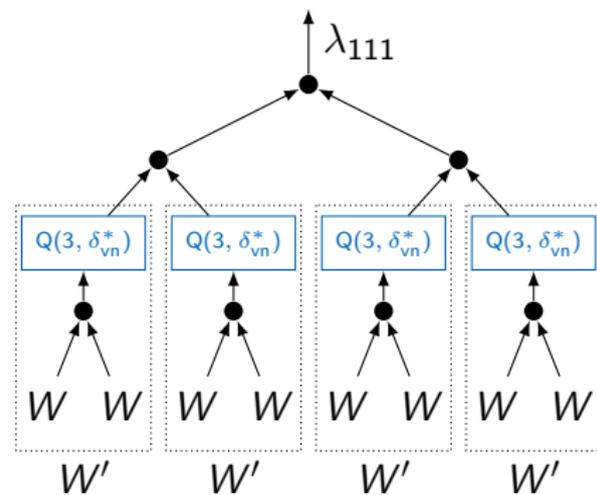


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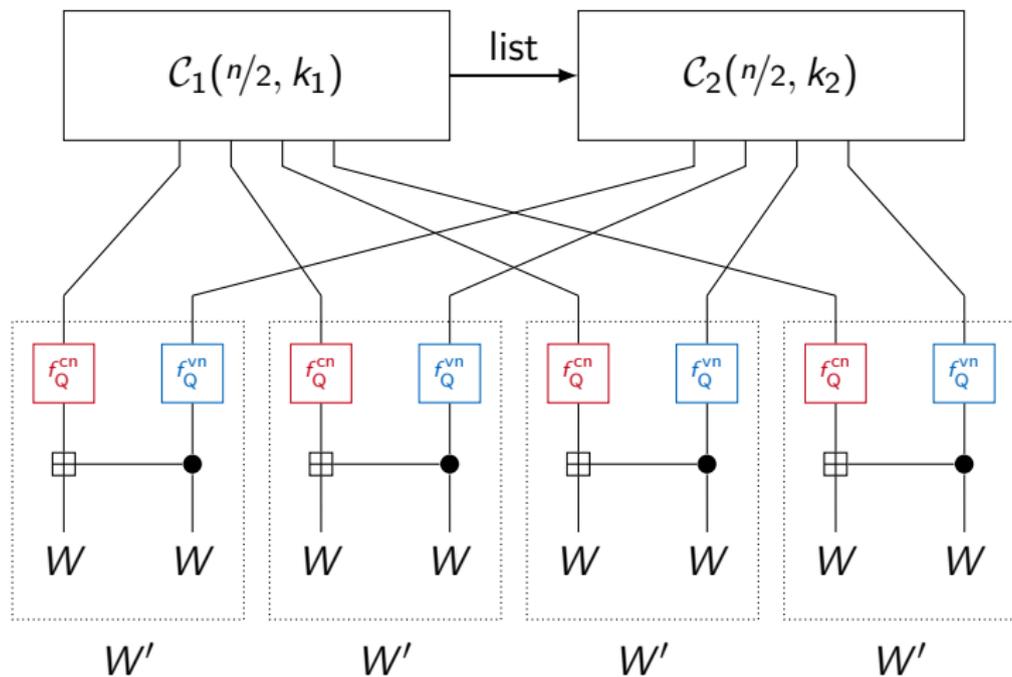
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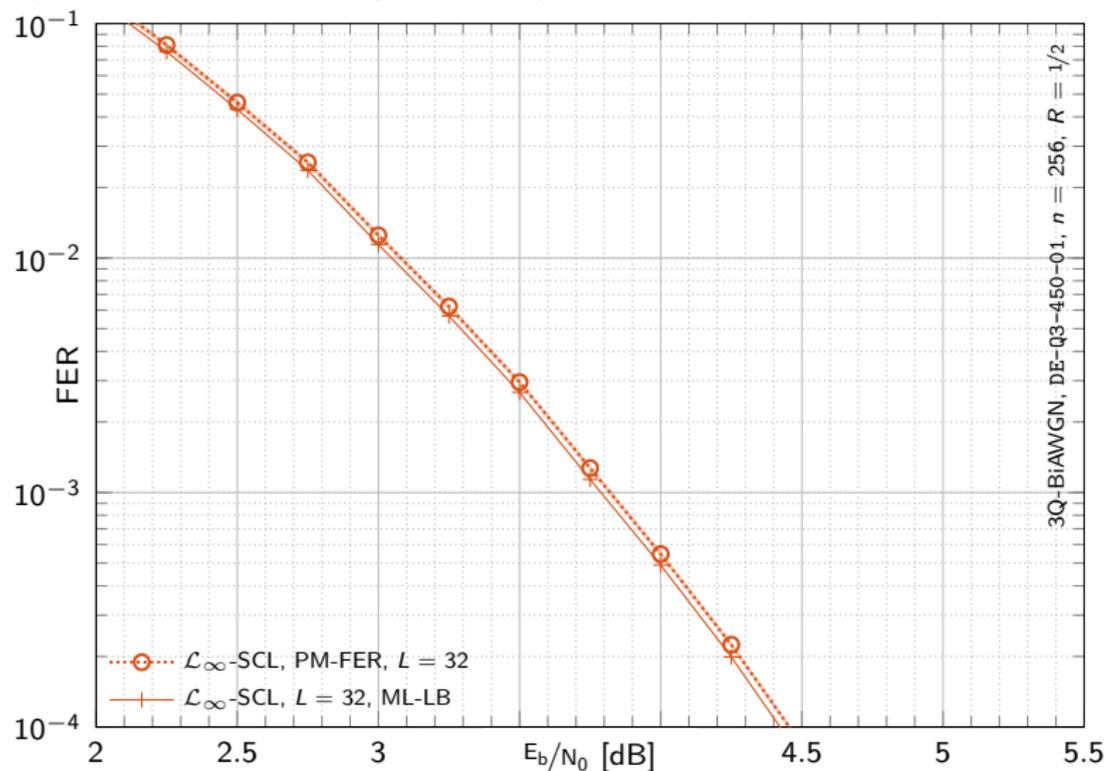
Variable nodes:



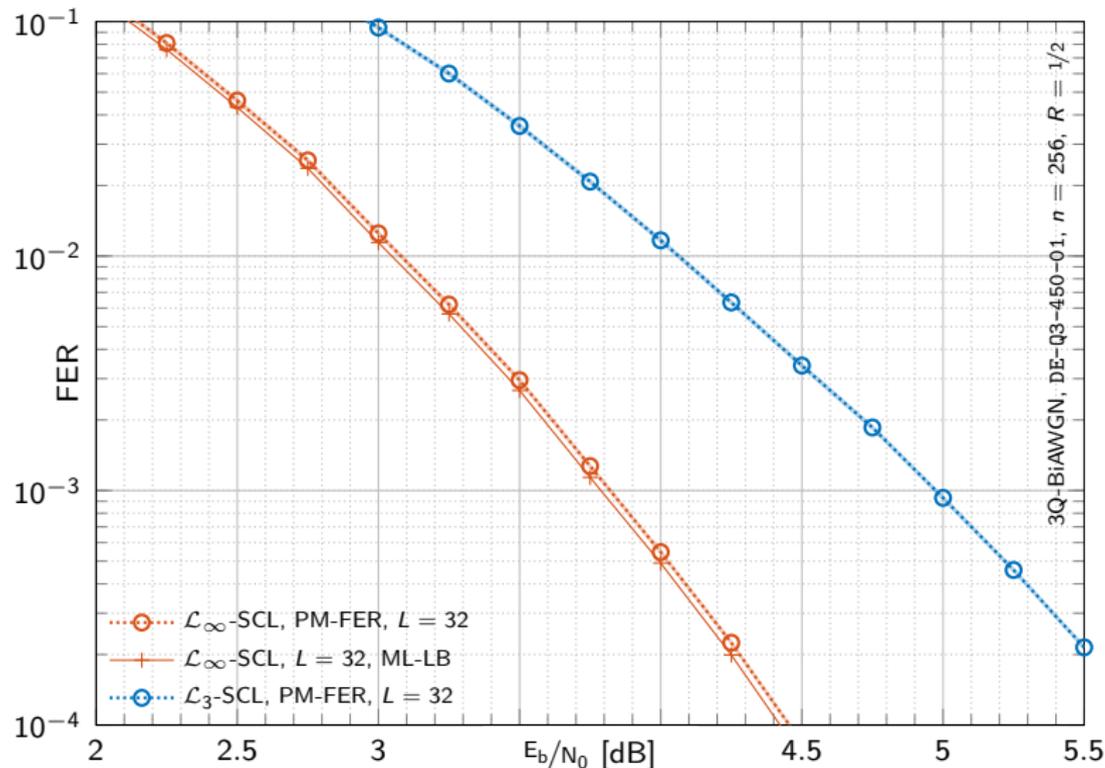
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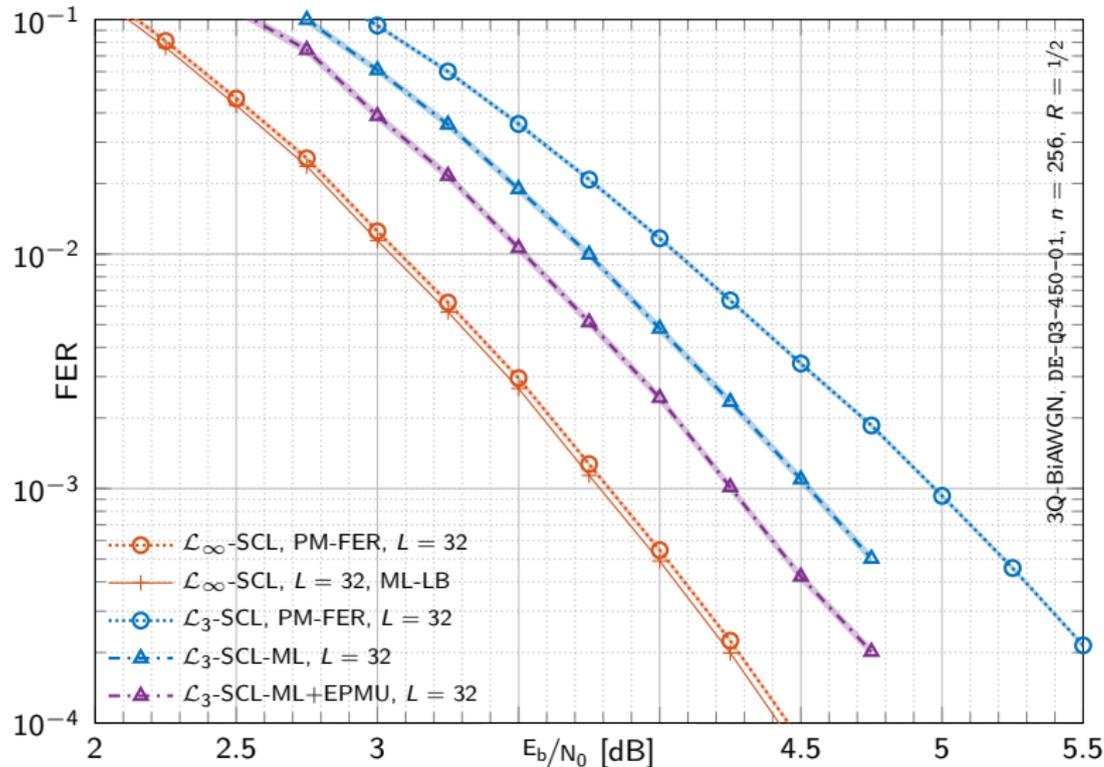
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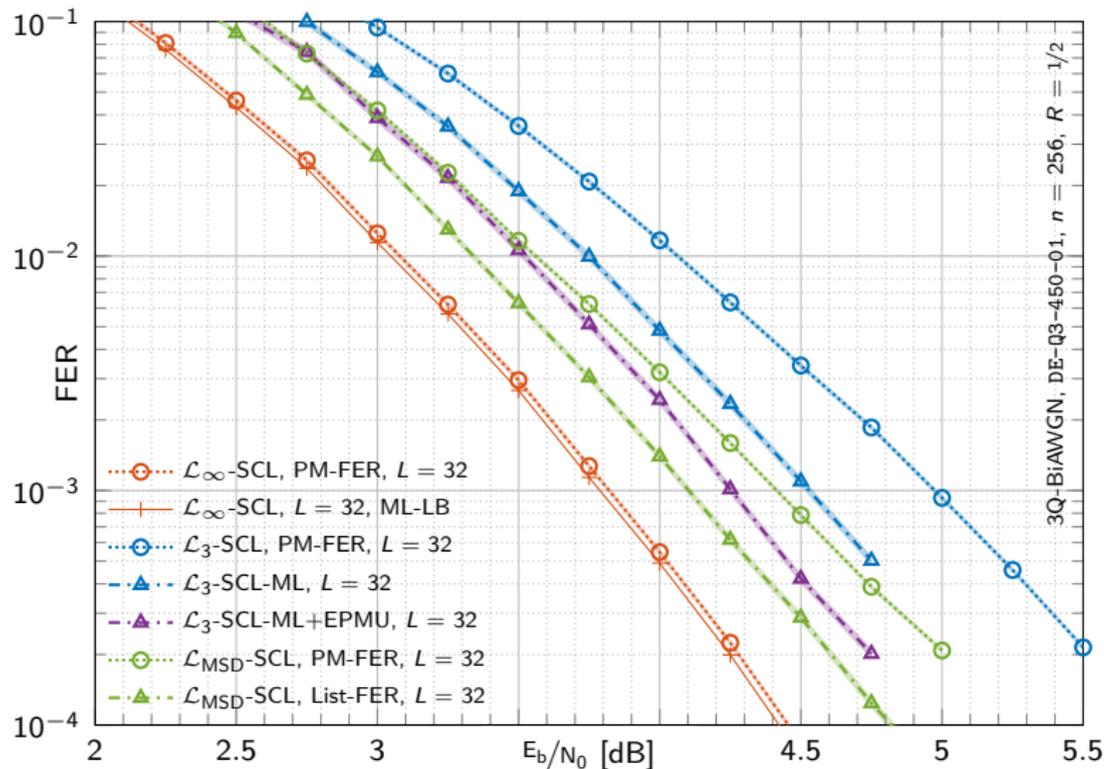
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- Interactions with outer codes (e.g., parity-checks)

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Thank you!

J. Neu, “Quantized Polar Code Decoders: Analysis and Design”, Master’s thesis at Technical University of Munich, September 2018, [arXiv:1902.10395](https://arxiv.org/abs/1902.10395)

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-  [E. Arkan](#). “Channel Polarization: A Method for Constructing Capacity-Achieving Codes for Symmetric Binary-Input Memoryless Channels”. In: *IEEE Trans. Inf. Theory* (2009).
-  [S. H. Hassani and R. Urbanke](#). “Polar Codes: Robustness of the Successive Cancellation Decoder with Respect to Quantization”. In: *Proc. IEEE Int. Symp. Inf. Theory (ISIT)*. 2012.
-  [A. Balatsoukas-Stimming, M. Bastani Parizi, and A. Burg](#). “LLR-Based Successive Cancellation List Decoding of Polar Codes”. In: *IEEE Trans. Signal Process.* (2015).

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