



# Neural Belief Propagation Decoding of CRC-Polar Concatenated Codes

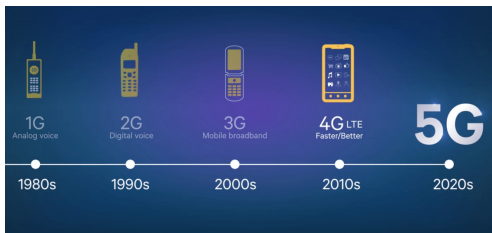
**Nghia Doan**<sup>1</sup>, Seyyed Ali Hashemi<sup>2</sup>, Elie Ngomseu Mambou<sup>1</sup>,  
Thibaud Tonnellier<sup>1</sup>, and Warren Gross<sup>1</sup>

<sup>1</sup>McGill University, Québec, Canada

<sup>2</sup>Stanford University, California, USA

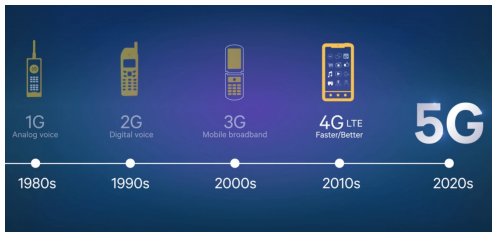
IEEE ICC  
Shanghai, China  
May 22, 2019

# Background



- ▶ Polar codes: selected for the eMBB control channel in 5G
- ▶ Cyclic redundancy check (CRC) is concatenated with polar codes in 5G for error detection
- ▶ Belief Propagation (BP): reasonable error-correction performance, **highly parallel**

# Background



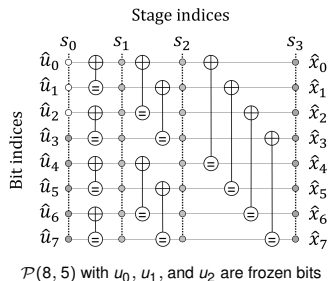
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- ▶ Belief Propagation (BP): reasonable error-correction performance, **highly parallel** → **high decoding throughput**

# Contribution

- ▶ Exploit the inherent CRC-polar concatenated factor graph to improve the error-correction performance under BP decoding
- ▶ Assign trainable weights to the concatenated factor graph to reduce the number of decoding iterations

# Polar codes

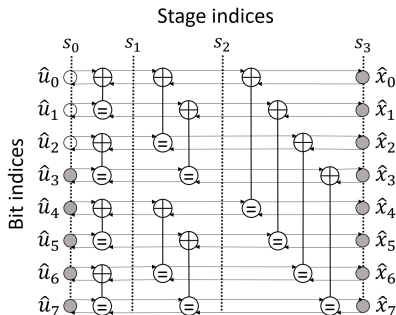
- ▶ Introduced by Arikan [1] in 2009
- ▶  $\mathcal{P}(N, K)$ ,  $N$ : code length,  $K$ : message length
- ▶ Code construction: based on polarization phenomenon
  - ▶  $K$  most reliable channels: information bits
  - ▶  $(N - K)$  least reliable channels: frozen bits



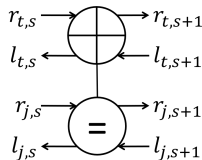
[1] E. Arikan, "Channel Polarization: A Method for Constructing Capacity-Achieving Codes for Symmetric Binary-Input Memoryless Channels", IEEE Trans. on Info. Theory, vol. 55, no. 7, pp. 3051–3073, July 2009.

# Belief Propagation (BP) Decoding

- ▶ Iterative message-passing algorithm
- ▶ Termination: CRC-based with predefined  $l_{\max}$  iterations
- ▶ Messages are calculated by Processing Elements (PEs)
- ▶ The message-passing operations can be unfolded



BP decoding for  $\mathcal{P}(8, 5)$

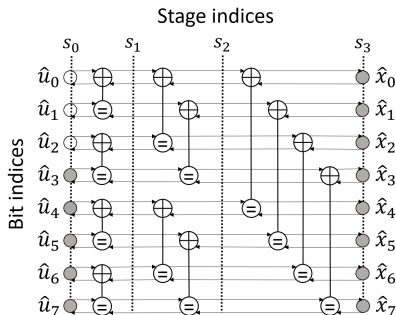


A polar PE

$l$ : right-to-left messages  
 $r$ : left-to-right messages  
 $0 \leq t, j \leq N - 1$ : bit indices  
 $0 \leq s \leq \log_2(N)$ : stage indices

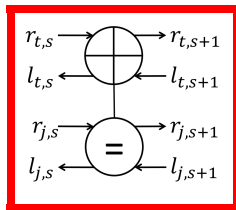
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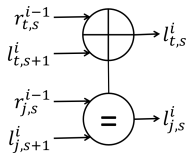
## Unfold



A polar PE

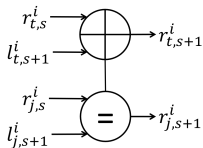
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# Belief Propagation (BP) Decoding



$$\begin{cases} l_{t,s}^i &= f(l_{t,s+1}^i, r_{j,s}^{i-1} + l_{j,s+1}^i) \\ l_{j,s}^i &= f(l_{t,s+1}^i, r_{t,s}^{i-1}) + l_{j,s+1}^i \end{cases}$$

A right-to-left polar PE



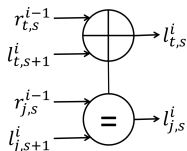
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A left-to-right polar PE

$i$ : iteration index,  $f(a, b) = \min(|a|, |b|) \operatorname{sgn}(a) \operatorname{sgn}(b)$

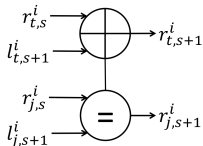


# Neural BP Decoding [2,3]



$$\begin{cases} l_{t,s}^i &= \mathbf{W}_0 f(l_{t,s+1}^i, \mathbf{W}_1 r_{j,s}^{i-1} + \mathbf{W}_2 l_{j,s+1}^i) \\ l_{j,s}^i &= \mathbf{W}_4 (\mathbf{W}_3 f(l_{t,s+1}^i, r_{t,s}^{i-1})) + \mathbf{W}_5 l_{j,s+1}^i \end{cases}$$

A right-to-left polar PE



$$\begin{cases} r_{t,s+1}^i &= \mathbf{W}_6 f(r_{t,s}^i, \mathbf{W}_7 l_{j,s+1}^i + \mathbf{W}_8 r_{j,s}^i) \\ r_{j,s+1}^i &= \mathbf{W}_{10} (\mathbf{W}_9 f(r_{t,s}^i, l_{t,s+1}^i)) + \mathbf{W}_{11} r_{j,s}^i \end{cases}$$

A left-to-right polar PE

$\mathbf{W}_m \in \mathbb{R}^+$  ( $0 \leq m \leq 11$ ): trainable weights

[2] E. Nachmani et al., "Deep learning methods for improved decoding of linear codes," IEEE J. of Sel. Topics in Signal Process., vol. 12, no. 1, pp. 119–131, February 2018.

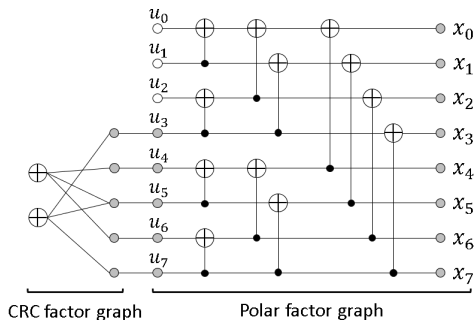
[3] W. Xu et al., "Improved polar decoder based on deep learning," in IEEE Int. Workshop on Signal Process. Syst., November 2017, pp. 1–6.

# CRC-Polar BP (CPBP) Decoding

- ▶ Exploit the CRC-Polar concatenated factor graph
- ▶ Run BP decoding on the CRC factor graph after  $I_{thr}$  iterations
- ▶ The choice of  $I_{thr}$  affects the error-correction performance

Algorithm: **CPBP**-( $I_{max}, I_{thr}$ )

```
1 for  $i = 0$  to  $I_{max}$  do  
2   Polar BP Decoding  
3   if  $i > I_{thr}$  then  
4     CRC BP Decoding  
5   if  $CRC == True$  then  
6     Output and Terminate
```

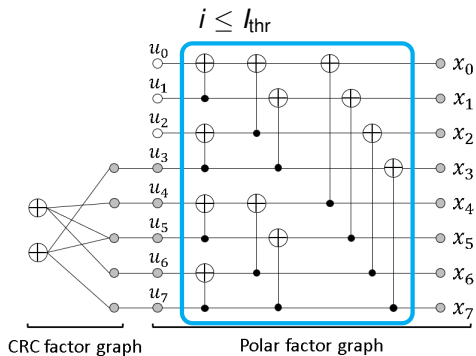


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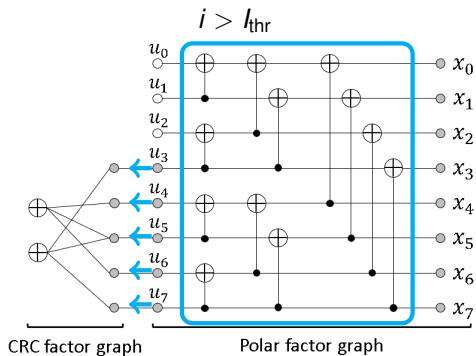


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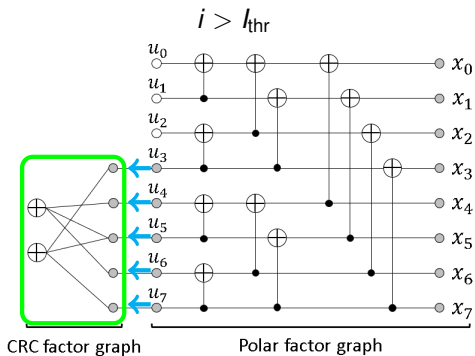


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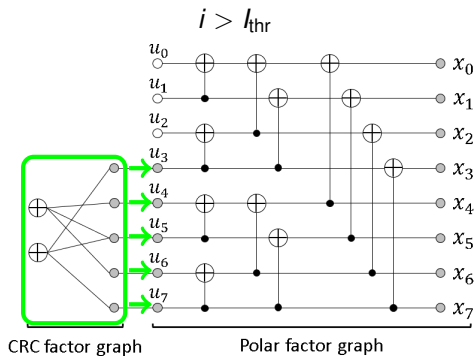


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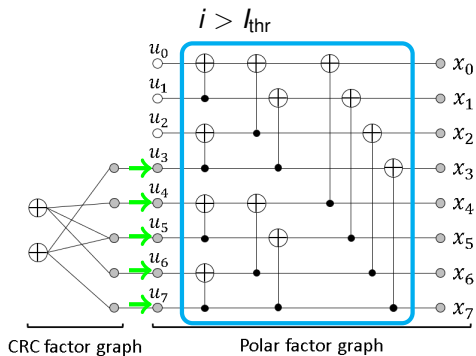


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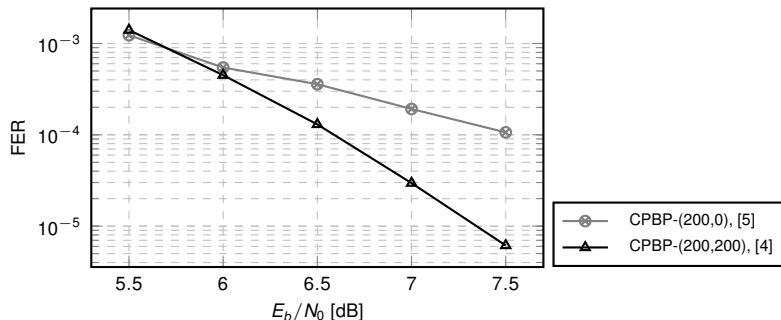
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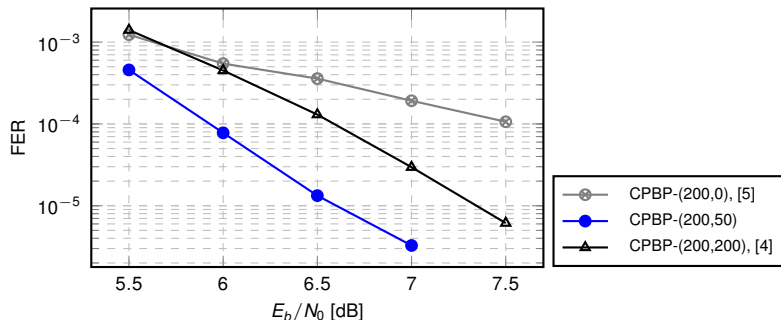
FER performance of various CPBP- $(l_{\max}, l_{\text{thr}})$  decoders for  $\mathcal{P}(128, 80)$  and a 16-bit CRC used in 5G.

[4] Y. Ren et al., "Efficient early termination schemes for belief-propagation decoding of polar codes," in IEEE 11th Int. Conf. on ASIC, Nov 2015, pp. 1–4.

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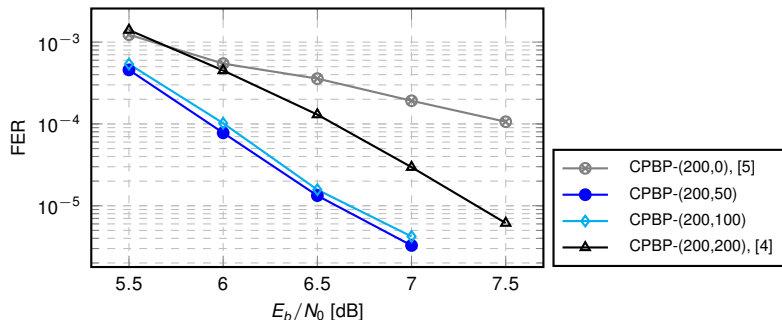


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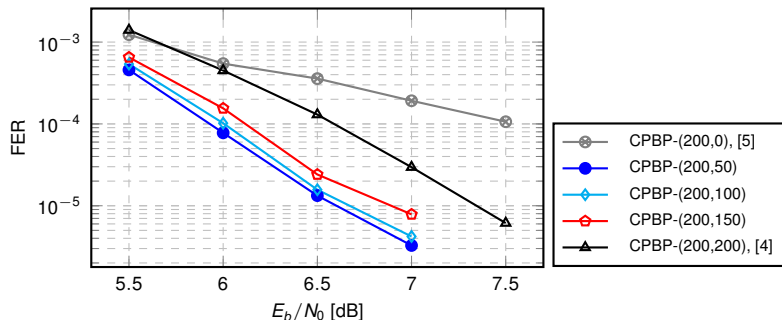


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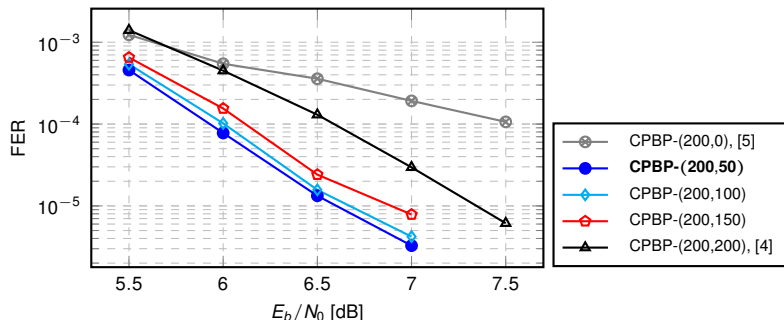


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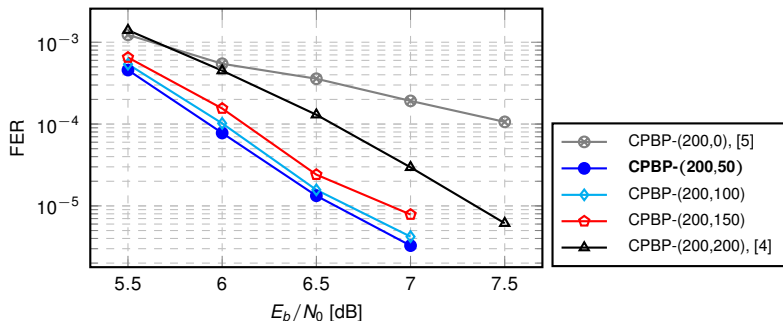
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- ▶ A small value of  $l_{\max}$  is needed for applications with strict latency requirements → **assign trainable weights to reduce  $l_{\max}$**

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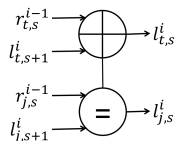
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# Neural CRC-Polar BP (NCPBP) Decoding

- ▶ Revise efficient weight assignment and sharing schemes for the CPBP decoder
- ▶ Preserve the symmetric property of the conventional BP decoding algorithm
- ▶ Training using all-zero codewords with stochastic gradient-descent based techniques

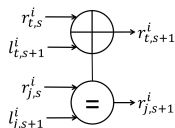
# Neural CRC-Polar BP (NCPBP) Decoding

## Weight assignment scheme



$$\begin{cases} l_{t,s}^i &= w_0 f(l_{t,s+1}^i, w_{1,2}(r_{j,s}^{i-1} + l_{j,s+1}^i)) \\ l_{j,s}^i &= w_{3,4} f(l_{t,s+1}^i, r_{t,s}^{i-1}) + w_5 l_{j,s+1}^i \end{cases}$$

A right-to-left polar PE



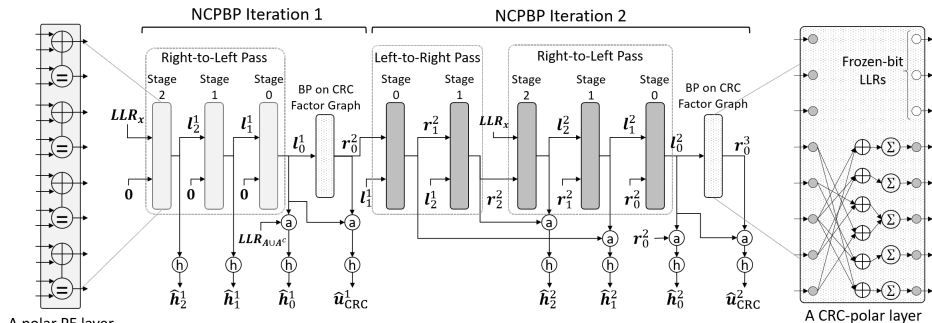
$$\begin{cases} r_{t,s+1}^i &= w_6 f(r_{t,s}^i, w_{7,8}(l_{j,s+1}^i + r_{j,s}^i)) \\ r_{j,s+1}^i &= w_{9,10} f(r_{t,s}^i, l_{t,s+1}^i) + w_{11} r_{j,s}^i \end{cases}$$

A left-to-right polar PE

$w_m \in \mathbb{R}^+$  are trainable weights.

# Neural CRC-Polar BP (NCPBP) Decoding

## Weight sharing scheme



A polar PE layer

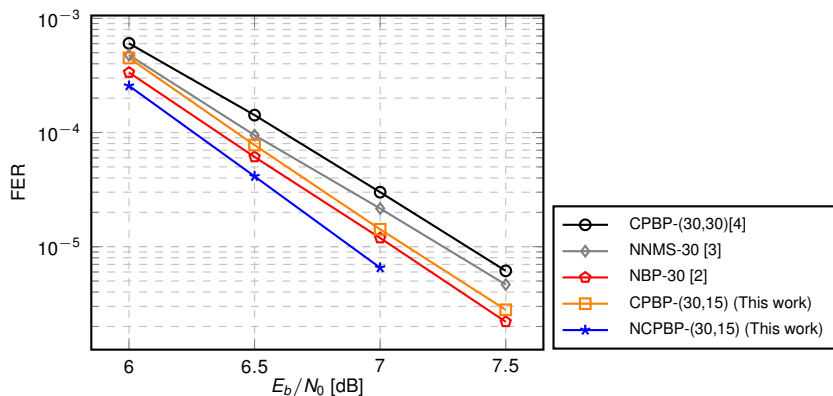
A CRC-polar layer

$$\begin{array}{c}
 x \\
 \searrow \\
 \textcircled{+} \\
 \swarrow \\
 y
 \end{array}
 \rightarrow \textcircled{a} \rightarrow x + y
 \quad
 \begin{array}{c}
 x \\
 \rightarrow \\
 \textcircled{h}
 \end{array}
 \rightarrow \begin{cases} \frac{1}{1 + e^x}, & \text{when training,} \\ \frac{1 - \text{sgn}(x)}{2}, & \text{when testing.} \end{cases}
 \quad
 x, y \in \mathbb{R}^N$$

- ▶ All polar PE layers in an iteration share the same weights
- ▶ All CRC-polar layers share the same set of weights



# Neural CRC-Polar BP (NCPBP) Decoding



FER performance of various BP decoders for  $\mathcal{P}(128, 80)$  and a 16-bit CRC used in 5G.

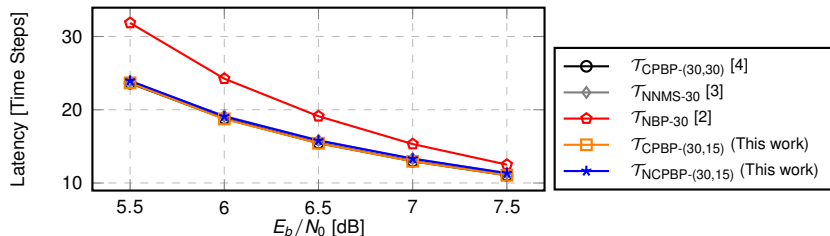
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# Neural CRC-Polar BP (NCPBP) Decoding

Average number of decoding time steps



- CRC-based early termination BP-based decoding:

$$\mathcal{T}_{\text{BP}} = (2n - 1)(l_{\text{ET}} - 1) + n$$

- Neural CPBP decoding:

$$\mathcal{T}_{\text{CPBP}} = \begin{cases} (2n - 1)(l_{\text{ET}} - 1) + n, & \text{if } l_{\text{ET}} \leq l_{\text{thr}} \\ (2n - 1)(l_{\text{ET}} - 1) + n + 2(l_{\text{ET}} - l_{\text{thr}}) & \text{otherwise.} \end{cases}$$

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# Neural CRC-Polar BP (NCPBP) Decoding

Number of weights required by different neural BP decoders.

Decoder	Number of weights
NBP-30 [2]	11520
NCPBP-(30,15) (This work)	8288
NNMS-30 [3]	3840

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# Conclusion

- ▶ We proposed CRC-polar BP (CPBP) decoding
- ▶ We proposed Neural CRC-polar BP (NCPBP) decoding with efficient weight assignment and sharing schemes
- ▶ For a 5G  $\mathcal{P}(128, 80)$  concatenated with a 16-bit CRC, NCPBP achieves up to 0.4 dB performance gain compared to state of the art, with almost no latency overhead

Thank You!