Neural Dynamic Successive-Cancellation Flip Decoding of Polar Codes

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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
- Successive Cancellation List (SCL) for CRC-Polar concatenated codes
 - Good error correction performance
 - Complexity increases when list size increases

Background



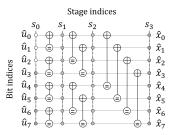
- Dynamic Successive Cancellation Flip (DSCF):
 - Comparable error-correction performance with SCL
 - Average decoding latency approaches that of SC at practical signal-to-noise ratio (SNR) regimes
 - Costly transcendental computations

This talk

- Propose a bit-flipping metric that only requires additions
- Propose a novel training framework for the decoder's parameter

Polar codes

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



 $\mathcal{P}(8,5)$ with $u_0, u_1,$ and u_2 are frozen bits

^[1] E. Arıkan, "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.

Successive Cancellation (SC) Decoding

Soft-value update rules:

$$\begin{cases} L_{s,i} = f(L_{s+1,i}, L_{s+1,i+2^s}) \\ L_{s,i+2^s} = g(L_{s+1,i}, L_{s+1,i+2^s}, \hat{\nu}_{s,i}), \end{cases}$$

where

$$\begin{cases} f(a,b) = \min(|a|,|b|) \operatorname{sgn}(a) \operatorname{sgn}(b) \\ g(a,b,c) = b + (1-2c)a, \end{cases}$$

Hard-value update rules:

$$\begin{cases} \hat{\nu}_{s+1,i} = \hat{\nu}_{s,i} \oplus \hat{\nu}_{s,i+2^s} \\ \hat{\nu}_{s+1,i+2^s} = \hat{\nu}_{s,i+2^s}, \end{cases}$$

where

$$\hat{u}_i = \hat{\nu}_{0,i} = egin{cases} 0 & \text{if } u_i \text{ is frozen,} \ rac{1-\mathsf{sgn}(L_{0,i})}{2} & \text{otherwise.} \end{cases}$$

^[1] E. Arıkan, "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.

Successive Cancellation Flip (SCF) Decoding

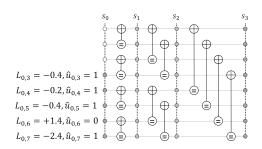
- Given that SC decoding fails the CRC verification
- SCF [2] flips the hard-decision of an information bit, then performs SC decoding again
- ▶ Flipping index $i = \arg \min_{\forall i \in \mathcal{A}} |L_{0,i}|$, where \mathcal{A} is the information set [2]

^[2] Orion, et al., "A low-complexity improved successive cancellation decoder for polar codes." IEEE Asilomar Conf. on Sign., Sys. and Comp, 2014

Successive Cancellation Flip (SCF) Decoding

Example:

- An all-zero codeword is sent
- Flipping position $i = \arg \min_{\forall i \in A} |L_{0,i}| = 4$

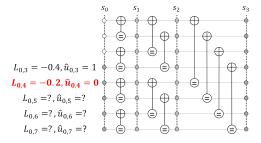


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Propose a bit-flipping metric for high-order error bits

[3] Chandesris et al., "Dynamic-SCFlip decoding of polar codes." IEEE Transactions on Communications 66.6 (2018): 2333-2345.

- Propose a bit-flipping metric for high-order error bits
- Bit-flipping probability at the i-th information bit [3]:

$$Pr_{i} = \prod_{\forall j \in \mathcal{E}_{\omega}, j \leq i} (1 - p_{j}^{*}) \prod_{\forall j \in \mathcal{A} \setminus \mathcal{E}_{\omega}, j < i} p_{j}^{*}$$
(1)

where

- \blacktriangleright \mathcal{E}_{ω} : the set of bit-flipping position at error order ω -th
- $ho_j^* = Pr(\hat{u}_i = u_i | \hat{u}_0^{i-1} = u_0^{i-1}) \approx \frac{1}{1 + \exp(-\alpha | L_{0,j} |)}$
- $ightharpoonup \alpha > 0$ is a perturbation parameter

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where

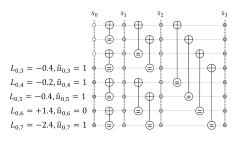
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- ► Flipping position $i = \arg \max_{\forall i \in A} Pr_i$

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Example:

- $ho_i^* = \frac{1}{1 + \exp(-\alpha |L_{0,i}|)}$
- $\rho = 0.3$
- \blacktriangleright At $\omega = 1$:

i	p_i^*	$1 - p_i^*$	Pr_i
3	0.53	0.47	-
4	0.51	0.49	-
5	0.53	0.47	-
6	0.60	0.40	-
7	0.67	0.33	-



ightharpoonup At $\omega=1$:

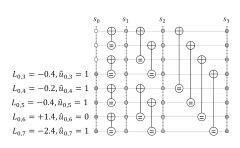
i	p_i^*	$1 - p_i^*$	Pri
3	0.53	0.47	0.47
4	0.51	0.49	0.26
5	0.53	0.47	0.13
6	0.60	0.40	0.06
7	0.67	0.33	0.03

$$Pr_3 = 1 - p_3^*$$

...

$$Pr_7 = \\ p_3^* \times p_4^* \times p_5^* \times p_6^* \times (1 - p_7^*)$$

 $ightharpoonup i = \operatorname{arg\,max}_{\forall i \in \mathcal{A}} Pr_i = 0$



ightharpoonup At $\omega = 1$:

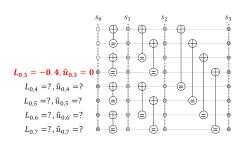
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 $ightharpoonup i = \operatorname{arg\,max}_{\forall i \in \mathcal{A}} Pr_i = 0$



ightharpoonup At $\omega=2$:

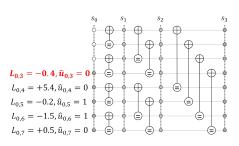
i	p_i^*	$1 - p_i^*$	Pr_i
3	0.53	0.47	0.47
4	0.83	0.17	0.08
5	0.51	0.49	0.19
6	0.61	0.39	0.07
7	0.46	0.54	0.06

$$Pr_4 = (1 - p_3^*) \times (1 - p_4^*)$$

...

►
$$Pr_7 = (1 - p_3^*) \times p_4^* \times p_5^* \times p_6^* \times (1 - p_7^*)$$

▶ $i = arg max_{\forall i \in A} Pr_i = 5$



ightharpoonup At $\omega=2$:

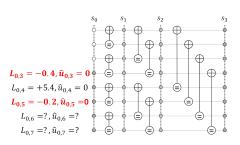
i	p_i^*	$1 - p_i^*$	Pri
3	0.53	0.47	0.47
4	0.83	0.17	0.08
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7	0.46	0.54	0.06

$$Pr_4 = (1 - p_3^*) \times (1 - p_4^*)$$

▶ ..

►
$$Pr_7 = (1 - p_3^*) \times p_4^* \times p_5^* \times p_6^* \times (1 - p_7^*)$$

▶ $i = arg max_{\forall i \in A} Pr_i = 5$



► To enable numerical stability

$$Q_{i} = -\frac{1}{\alpha} \ln(Pr_{i}) = \sum_{\substack{\forall j \in \mathcal{A} \\ j \leq i}} \frac{1}{\alpha} \ln\left(1 + \exp\left(-\alpha |L_{0,j}|\right)\right) + \sum_{\substack{\forall j \in \mathcal{E}_{\omega} \\ j < i}} |L_{0,j}|.$$

$$(3)$$

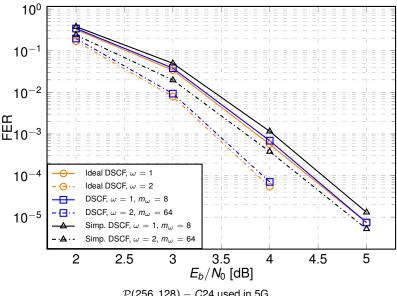
► Computing *Q_i* requires costly In and exp functions!

Simplify Q_i using a conventional technique [4]:

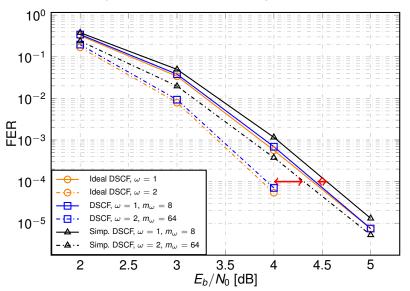
$$ln(1 + exp(x)) \approx ReLU(x)$$
 (4)

$$\tilde{Q}_{i} = \sum_{\substack{\forall j \in \mathcal{A} \\ j \leq i}} \frac{1}{\alpha} \operatorname{ReLU} \left(-\alpha | L_{0,j} | \right) + \sum_{\substack{\forall j \in \mathcal{E}_{\omega} \\ j \leq i}} |L_{0,j}| = \sum_{\substack{\forall j \in \mathcal{E}_{\omega} \\ j \leq i}} |L_{0,j}| \tag{5}$$

^[4] Balatsoukas-Stimming et al., "LLR-based successive cancellation list decoding of polar codes." IEEE Trans. on Sig. Process. 63.19 (2015): 5165-5179.



 $\mathcal{P}(256, 128) - C24$ used in 5G.



Simplifying DSCF hurts the FER performance!

▶ Introduce an additive perturbation parameter to estimate p_i^*

$$\rho_j^* \approx \frac{1}{1 + \exp(\beta - |L_{0,j}|)} \tag{6}$$

instead of [3]

$$p_j^* pprox rac{1}{1 + \exp(-lpha |L_{0,j}|)}$$

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Introduce an additive perturbation parameter to estimate p_i*

$$p_j^* \approx \frac{1}{1 + \exp(\beta - |L_{0,j}|)} \tag{7}$$

The tailored bit-flipping metric under NDSCF decoding

$$Q_{i}^{*} = \sum_{\substack{\forall j \in \mathcal{A} \\ j \leq i}} \text{ReLU}\left(\beta - |L_{0,j}|\right) + \sum_{\substack{\forall j \in \mathcal{E}_{\omega} \\ j \leq i}} |L_{0,j}| \tag{8}$$

► DSCF

$$Q_i = \sum_{\substack{\forall j \in \mathcal{A} \\ j \leq i}} \frac{1}{\alpha} \ln \left(1 + \exp \left(-\alpha |L_{0,j}| \right) \right) + \sum_{\substack{\forall j \in \mathcal{E}_{\omega} \\ j \leq i}} |L_{0,j}|.$$

Simplified DSCF

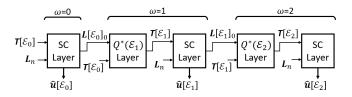
$$\tilde{Q}_i = \sum_{\substack{\forall j \in \mathcal{E}_{\omega} \ j < i}} |L_{0,j}|.$$

NDSCF (proposed)

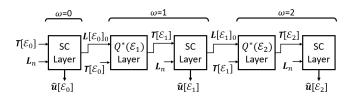
$$Q_{i}^{*} = \sum_{\substack{orall j \in \mathcal{A} \ j < i}} \operatorname{ReLU}\left(eta - |L_{0,j}|
ight) + \sum_{\substack{orall j \in \mathcal{E}_{\omega} \ j < i}} |L_{0,j}|$$

Decoder	+	×	In, exp
DSCF	✓	✓	✓
Simplified DSCF	\checkmark		
NDSCF (proposed)	✓		

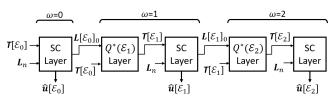
The optimization of β can be done using deep-learning techniques



- SC layer: SC decoding
- $Q^*(\mathcal{E}_{\omega})$: bit-flipping metric layer at error order ω -th
- ► $T[\mathcal{E}_{\omega}]$: bit-flipping vector at error order ω -th



- ► The proposed metric Q_i^{*} is independent from u, hence all-zero codewords are used for training
- SC decoding can be implemented as a network layer to avoid the labeling task for the correct flipping position



Mapping from Q^* to T^* :

▶ Inference:

$$m{Q}^* = \{Q_0^*, \dots, Q_{i_{min}}^*, \dots, Q_n^*\} \ ext{where} \ Q_{i_{min}}^* = \min\{m{Q}^*\}, \ ext{thus} \ m{T}^* = \{T_0^* = 1, \dots, T_{i_{min}}^* = -1, \dots, T_n^* = 1\}.$$

Training:

Let
$$Q_{l_{2ndmin}}^* = \min\{ oldsymbol{Q}^* \setminus Q_{l_{min}}^* \}$$
 and $Q_{ au} = rac{Q_{l_{2ndmin}}^* + Q_{l_{min}}^*}{2}$, $oldsymbol{T}^* = \{ \mathcal{T}_0^* = \tanh(Q_0^* - Q_{ au}), \ldots, \mathcal{T}_n^* = \tanh(Q_n^* - Q_{ au}) \}.$

► Training loss:

$$\lambda = \sum_{\omega=1}^{2} I_{\omega},\tag{9}$$

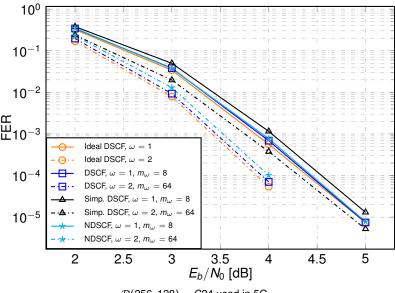
where

$$\begin{split} I_{\omega} &= \begin{cases} (\hat{\boldsymbol{u}}[\mathcal{E}_{\omega}]_{t_{\omega}} - \boldsymbol{u}_{t_{\omega}})^{2} & \text{if CRC}(\hat{\boldsymbol{u}}[\mathcal{E}_{\omega}]) \neq 0, \, i_{\omega}^{*} \geq t_{\omega}, \\ 0 & \text{otherwise,} \end{cases} \\ &\approx \begin{cases} \frac{1}{\left(1 + \exp\left(\mathcal{L}[\mathcal{E}_{\omega}]_{0,t_{\omega}}\right)\right)^{2}} & \text{if CRC}(\hat{\boldsymbol{u}}[\mathcal{E}_{\omega}]) \neq 0, \, i_{\omega}^{*} \geq t_{\omega}, \\ 0 & \text{otherwise,} \end{cases} \end{split}$$

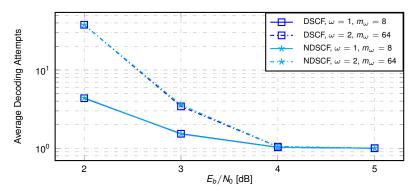
 t_{ω} : the first erroneous bit position of $\hat{\pmb{u}}[\mathcal{E}_{\omega}]$ $u_{t_{\omega}}=0$

Optimized parameters:

ω	1	2
β	2.801	2.196



P(256, 128) - C24 used in 5G.



Average decoding attempts for $\mathcal{P}(256, 128) - \textit{C24}$ used in 5G.

Conclusion

- Propose a simplified DSCF decoding algorithm using neural-network techniques
- Only additions are used to calculate the bit-flipping metric
- Incur negligible error-correction performance loss compared to DSCF decoding
- Preserve the same average decoding latency compared to SC decoding

Thank You!