

Error-Correcting Codes for Nanopore Sequencing

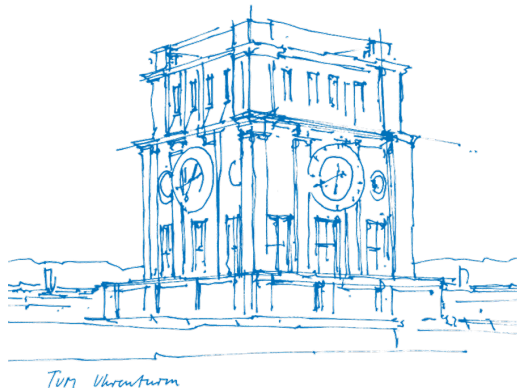
Anisha Banerjee¹

Joint work with
Yonatan Yehezkeally¹, Antonia Wachter-Zeh¹,
and Eitan Yaakobi²

¹Technical University of Munich
Institute for Communications Engineering

²Technion – Israel Institute of Technology
Department of Computer Science

June 26, 2023



Outline

Introduction

Channel Model

Minimum Redundancy

Error-correcting Code

Conclusion

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- Need for dense, reliable, robust storage media
 - ▶ Molecular storage paradigms *e.g.*, *DNA storage*.¹



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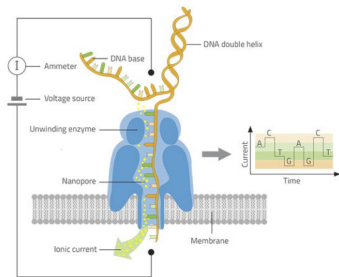


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- Nanopore sequencing²:
 - + Can read longer DNA strands
 - + More portable
 - + Low cost
 - High error rates



Source: "Decoding DNA with a pocket-sized sequencer," *Science in School*. <https://www.scienceinschool.org/article/2018/decoding-dna-pocket-sized-sequencer/>

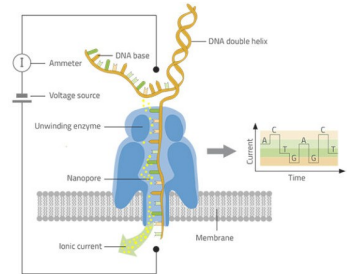
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 - High error rates
- *Aim*: Design coding techniques tailored for nanopore sequencing!

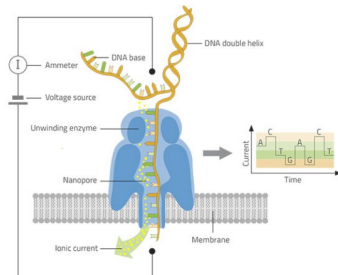


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Nanopore Sequencing



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- Sources of noise [MDK18] :
 - ▶ Nanopore holds $\ell > 1$ nucleotides at a time \rightarrow Intersymbol interference (ISI)!
 - ▶ Strand moves irregularly \rightarrow backtracking & skipping.
 - ▶ Noisy measurements

[MDK18] W. Mao *et al.*, "Models and information-theoretic bounds for nanopore sequencing," *IEEE Trans. Inf. Theory*, vol. 64, no. 4, pp. 3216–3236, Apr. 2018

Prior Work

- [MDK18] → Introduced a mathematical model.
→ Established bounds on capacity.
- [HCW21] → Algorithm to compute capacity of an abstracted, deterministic channel model.
→ Coding schemes.
- [MVS22] → Finite-state Markov channel model incorporating major noise sources.
→ Generalized MAP detection algorithms.
- [CVVY21] considered a similar channel for racetrack memories:
→ Computed information limits.
→ Proposed error-correcting codes.

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[HCW21] R. Hulett *et al.*, “On Coding for an Abstracted Nanopore Channel for DNA Storage,” in *IEEE Intl. Symp. on Inf. Theory (ISIT)*, Melbourne, Australia, Jul. 2021, pp. 2465–2470

[MVS22] B. McBain *et al.*, “Finite-state semi-markov channels for nanopore sequencing,” in *IEEE Intl. Symp. Inf. Theory (ISIT)*, Espoo, Finland, Jun. 2022, pp. 216–221

[CVVY21] Y. M. Chee *et al.*, “Coding for transverse-reads in domain wall memories,” in *IEEE Intl. Symp. Inf. Theory (ISIT)*, Melbourne, Australia, Jul. 2021, pp. 2924–2929

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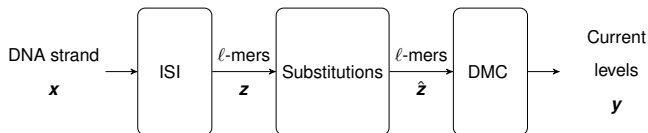
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Minimum Redundancy

Error-correcting Code

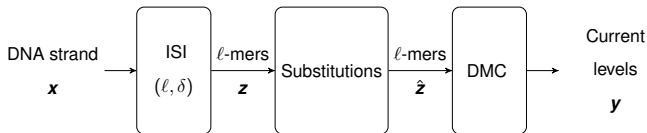
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Simplified Model



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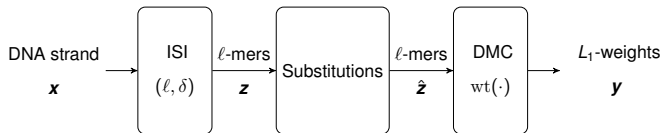


We model

- ISI as (ℓ, δ) -sliding window
- Measurement noise as substitutions

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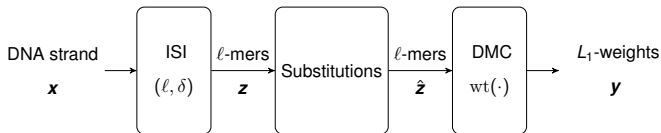
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Assumptions:

- Input $\mathbf{x} \in \{0, 1\}^n$
- Output: L_1 -weights of ℓ -mers

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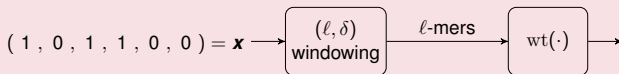
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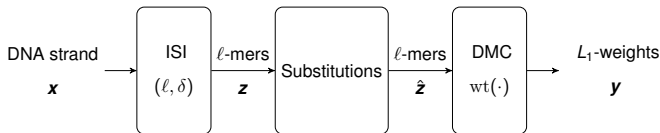
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Example $(\ell = 3, \delta = 1)$



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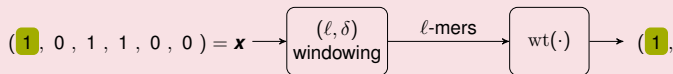
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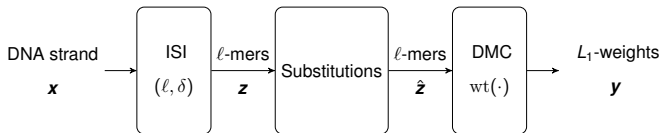
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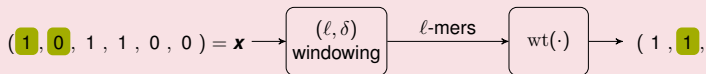
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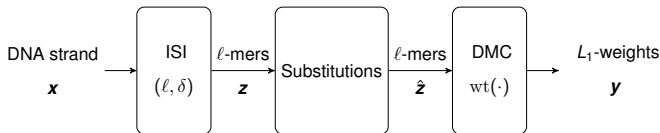
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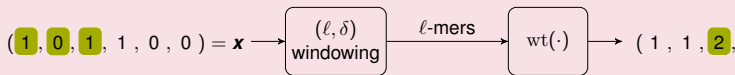
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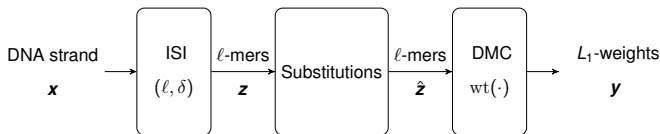
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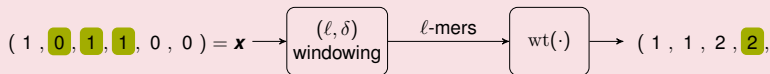
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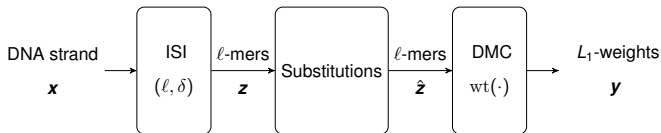
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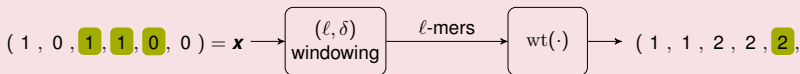
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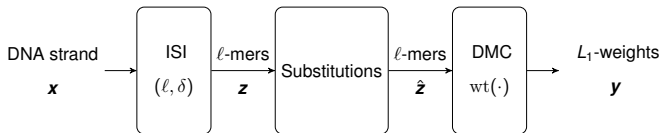
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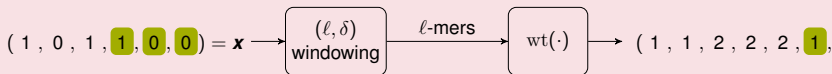
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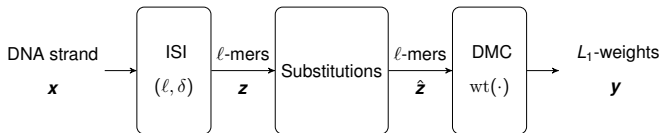
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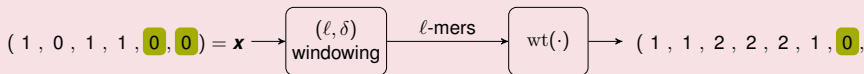
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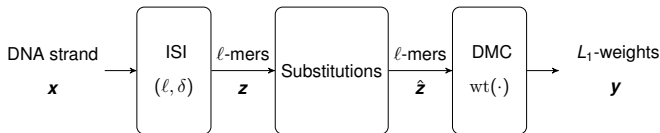
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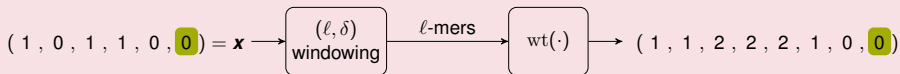
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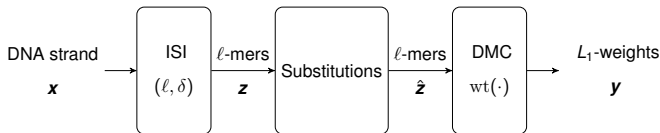
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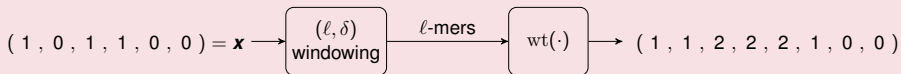
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Definition

The (ℓ, δ) -read vector of any $\mathbf{x} \in \Sigma_2^n$ is

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where for $i \notin [n]$, let $x_i = 0$ and $\text{wt}(\cdot)$ denotes L_1 -weight.

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Our Results

- $\log \log n - o(1)$ min redundancy to correct 1 substitution in $(\ell \geq 3, \delta = 1)$ -read vectors
- Introduce a 1-substitution correcting code that is optimal up to a constant.

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Read Vector

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- Read vectors have special properties.
 - ▶ $|\mathcal{R}(\mathbf{x})_i - \mathcal{R}(\mathbf{x})_{i-1}| \leq 1 \iff \mathcal{R}(\mathbf{x})_i - \mathcal{R}(\mathbf{x})_{i-1} = x_i - x_{i-\ell}$

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- Can correct errors of abs. value ≥ 2 .
 - ▶ Suffices to **consider ± 1 errors**.

Error-correcting Code

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\mathcal{C} is a t -substitution $(\ell, 1)$ -read code if for all $\mathbf{x}, \mathbf{y} \in \mathcal{C}$,

$$d_H(\mathcal{R}(\mathbf{x}), \mathcal{R}(\mathbf{y})) > 2t,$$

where $d_H(\cdot, \cdot)$ denotes Hamming distance.

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 - ▶ Minimum redundancy required?

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Definition

\mathcal{C} is a t -substitution $(\ell, 1)$ -read code if for all $\mathbf{x}, \mathbf{y} \in \mathcal{C}$,

$$d_H(\mathcal{R}(\mathbf{x}), \mathcal{R}(\mathbf{y})) > 2t,$$

where $d_H(\cdot, \cdot)$ denotes Hamming distance.

- *Aim:* Find a 1-substitution $(\ell, 1)$ -read code.
 - ▶ Minimum redundancy required?
 - ▶ When does $d_H(\mathcal{R}(\mathbf{x}), \mathcal{R}(\mathbf{y})) \leq 2$ occur?

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Confusable Read Vectors

- When does $d_H(\mathcal{R}(\mathbf{x}), \mathcal{R}(\mathbf{y})) \leq 2$ occur?

Lemma

For $\ell > 1$, $\delta = 1$ and any distinct $\mathbf{x}, \mathbf{y} \in \{0, 1\}^n$, $d_H(\mathcal{R}(\mathbf{x}), \mathcal{R}(\mathbf{y})) \neq 1$.

Confusable Read Vectors

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For $\ell > 1$, $\delta = 1$ and any distinct $\mathbf{x}, \mathbf{y} \in \{0, 1\}^n$, $d_H(\mathcal{R}(\mathbf{x}), \mathcal{R}(\mathbf{y})) \neq 1$.

Proof idea:

- Use read-vector properties:
 - ▶ $\sum_i \mathcal{R}_i \bmod \ell = 0$.
 - ▶ $|\mathcal{R}_{i+1} - \mathcal{R}_i| \leq 1$.

Confusable Read Vectors

- When does $d_H(\mathcal{R}(\mathbf{x}), \mathcal{R}(\mathbf{y})) = 2$ occur?

Confusable Read Vectors

- When does $d_H(\mathcal{R}(\mathbf{x}), \mathcal{R}(\mathbf{y})) = 2$ occur?

Theorem

For $\ell \geq 3$ and any $\mathbf{x}, \mathbf{y} \in \Sigma_2^n$, the following are equivalent:

- $d_H(\mathcal{R}(\mathbf{x}), \mathcal{R}(\mathbf{y})) = 2$.

Example ($\ell = 3, \delta = 1$)

$$\mathbf{x} = (1, 0, 1, 1, 0, 0), \quad \mathcal{R}(\mathbf{x}) = (\textcolor{blue}{1}, 1, 2, 2, 2, 1, \textcolor{blue}{0}, 0)$$

$$\mathbf{y} = (0, 1, 1, 0, 1, 0), \quad \mathcal{R}(\mathbf{y}) = (\textcolor{blue}{0}, 1, 2, 2, 2, 1, \textcolor{blue}{1}, 0)$$

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- There exist $p \geq 1$, $i \in [n - (p - 1)\ell - 1]$ s.t. $\forall m \in \Sigma_p$, $\mathbf{x}_{i+ml}^{i+ml+1} = (1, 0)$, $\mathbf{y}_{i+ml}^{i+ml+1} = (0, 1)$
 (or vice versa), and $x_r = y_r$ for all $r \notin \bigcup_{m \in \Sigma_p} \{i + m\ell, i + m\ell + 1\}$.

Further, if these conditions hold, then $j = i + p\ell$ in the above notation.

Example ($\ell = 3, \delta = 1$)

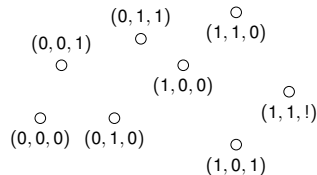
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Minimum Redundancy

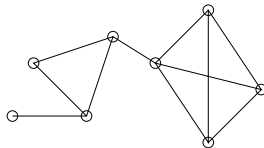
- Consider graph with all vertices in $\{0, 1\}^n$.



[K94] D. E. Knuth, "The sandwich theorem," *The Electronic Journal of Combinatorics*, vol. 1, no. 1, A1, Apr. 1994
 [CKY22] J. Chrisnata *et al.*, "Correcting deletions with multiple reads," *IEEE Trans. Inf. Theory*, vol. 68, no. 11, pp. 7141–7158, Nov. 2022

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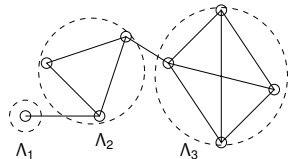
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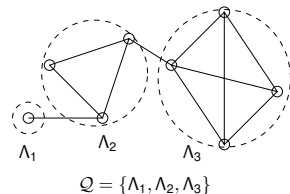
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 - Clique $\Lambda \subset \{0, 1\}^n \rightarrow$ all vertices pairwise-adjacent.



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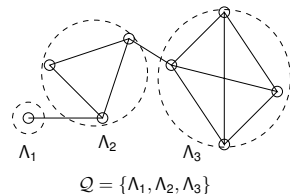


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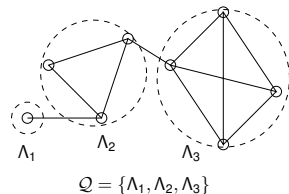


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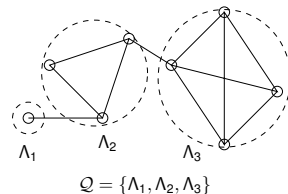


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Theorem

The redundancy of a 1-substitution $(\ell, 1)$ -read code is bounded from below by

$$\log_2 \log_2(n) - o(1).$$

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$$\mathcal{R}(\mathbf{x}) = (\text{wt}(x_1), \text{wt}(\mathbf{x}_1^2), \dots, \text{wt}(\mathbf{x}_1^\ell), \text{wt}(\mathbf{x}_2^{\ell+1}), \dots, \text{wt}(\mathbf{x}_{n-\ell+1}^n), \dots, \text{wt}(\mathbf{x}_{n-1}^n), \text{wt}(x_n))$$

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$$(0, \text{wt}(\mathbf{x}_1), \text{wt}(\mathbf{x}_1^2), \dots, \text{wt}(\mathbf{x}_1^\ell), \text{wt}(\mathbf{x}_2^{\ell+1}), \dots, \text{wt}(\mathbf{x}_{n-\ell+1}^n), \dots, \text{wt}(\mathbf{x}_{n-1}^n), \text{wt}(\mathbf{x}_n), 0)$$

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Derivative

$$(x_1, x_2, \dots, x_\ell, x_{\ell+1} - x_1, x_{\ell+2} - x_2, \dots, x_n - x_{n-\ell}, -x_{n-\ell+1}, \dots, -x_{n-1}, -x_n)$$

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↓ Sub-sequence

$$(x_1, x_{\ell+1} - x_1, x_{2\ell+1} - x_{\ell+1}, \dots, -x_{\lfloor \frac{n-1}{\ell} \rfloor \ell + 1})$$

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Sub-sequence

$$(\mathbf{x}_1, \mathbf{x}_{\ell+1} - \mathbf{x}_1, \mathbf{x}_{2\ell+1} - \mathbf{x}_{\ell+1}, \dots, -\mathbf{x}_{\lfloor \frac{n-1}{\ell} \rfloor \ell + 1})$$

- Call this **read sub-derivative** $\Delta^0(\mathbf{x})$.

Reconstruction from Read Vectors

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Sub-sequence

$$(x_{\alpha+1}, x_{\ell+\alpha+1} - x_{\alpha+1}, x_{2\ell+\alpha+1} - x_{\ell+\alpha+1}, \dots, -x_{\lfloor \frac{n-\alpha-1}{\ell} \rfloor \ell + \alpha + 1}) \quad [\text{Sums to 0}]$$

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- Call this **read sub-derivative** $\Delta^\alpha(\mathbf{x})$, $\alpha \in \{0, \dots, \ell - 1\}$.
- Helps reconstruct subsequence $(x_{\alpha+1}, x_{\ell+\alpha+1}, x_{2\ell+\alpha+1}, \dots)$.

Read Sub-derivative

Sub-derivative of read vector

For any $\alpha \in \{0, \dots, \ell - 1\}$, the α -th read sub-derivative of $\mathcal{R}(\mathbf{x})$ is

$$\begin{aligned}\Delta^\alpha(\mathbf{x}) &= (\mathcal{R}(\mathbf{x})_{\alpha+1} - \mathcal{R}(\mathbf{x})_\alpha, \mathcal{R}(\mathbf{x})_{\alpha+\ell+1} - \mathcal{R}(\mathbf{x})_{\alpha+\ell}, \dots, \mathcal{R}(\mathbf{x})_{\alpha+k\ell+1} - \mathcal{R}(\mathbf{x})_{\alpha+k\ell}) \\ &= (x_{\alpha+1}, x_{\alpha+\ell+1} - x_{\alpha+1}, \dots, x_{\alpha+(k-1)\ell+1} - x_{\alpha+(k-2)\ell+1}, -x_{\alpha+(k-1)\ell+1}),\end{aligned}$$

It holds that $\sum_i \Delta^\alpha(\mathbf{x})_i = 0$.

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$$\begin{aligned}\mathcal{R}(\mathbf{x})'_i - \mathcal{R}(\mathbf{x})_i &= \sum_k \Delta^{i-1 \bmod \ell}(\mathbf{x})'_k \\ &= - \sum_k \Delta^{i \bmod \ell}(\mathbf{x})'_k \neq 0\end{aligned}$$

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Takeaway

Given $\mathcal{R}(\mathbf{x})'$, can determine **error value** & **position upto mod ℓ** .

Reconstruction with Noisy Sub-derivative

- Can an erroneous read sub-derivative reveal region of error?

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Example ($\ell = 3, \delta = 1$)

$$\mathcal{R}(\mathbf{x}) \xrightarrow{\text{error at } i_1} \mathcal{R}(\mathbf{x})' = (1, 1, 2, 3, 2, 1, 0, 0)$$

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$$\begin{aligned} \Delta^0(\mathbf{x})' &= (\mathcal{R}(\mathbf{x})'_1, \mathcal{R}(\mathbf{x})'_4 - \mathcal{R}(\mathbf{x})'_1, \mathcal{R}(\mathbf{x})'_7 - \mathcal{R}(\mathbf{x})'_6) \\ &= (x'_1, \quad x'_4 - x'_1, \quad -x'_4) \\ &= (1, \quad 1, \quad -1) \end{aligned}$$

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$$\mathcal{R}(\mathbf{x}) \xrightarrow{\text{error at } i_1} \mathcal{R}(\mathbf{x})' = (1, 1, 2, 3, 2, 1, 0, 0)$$

$$\bullet \sum_k \Delta^0(\mathbf{x})'_k = -\sum_k \Delta^1(\mathbf{x})'_k = 1 \implies i_1 \bmod \ell = 1 \implies i_1 \in \{1, 4, 7\}$$

$$\begin{aligned} \Delta^0(\mathbf{x})' &= (\mathcal{R}(\mathbf{x})'_1, \mathcal{R}(\mathbf{x})'_4 - \mathcal{R}(\mathbf{x})'_1, \mathcal{R}(\mathbf{x})'_7 - \mathcal{R}(\mathbf{x})'_6) \\ &= (x'_1, x'_4 - x'_1, -x'_4) \\ &= (1, 1, -1) \end{aligned}$$

- Left to right reconstruction:

$$\blacktriangleright x'_1 = 1.$$

Reconstruction with Noisy Sub-derivative

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$$\blacktriangleright x'_1 = 1.$$

$$\blacktriangleright x'_4 - x'_1 = 1 \implies x'_4 = 2 \notin \{0, 1\} \text{ ✗}$$

Reconstruction with Noisy Sub-derivative

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- Left to right reconstruction:
 - ▶ $x'_1 = 1$.
 - ▶ $x'_4 - x'_1 = 1 \implies x'_4 = 2 \notin \{0, 1\}$ ✗
- Right to left reconstruction:
 - ▶ $x'_4 = 1$.

Reconstruction with Noisy Sub-derivative

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- Left to right reconstruction:
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 - ▶ $x'_4 - x'_1 = 1 \implies x'_4 = 2 \notin \{0, 1\} \mathbf{x}$
- Right to left reconstruction:
 - ▶ $x'_4 = 1$.
 - ▶ $x'_4 - x'_1 = 1 \implies x'_1 = 0$

Reconstruction with Noisy Sub-derivative

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Reconstruction with Noisy Sub-derivative

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- Right to left reconstruction:
 - ▶ $x'_4 = 1$.
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 - ▶ $x'_1 = 1$ ✗

Reconstruction with Noisy Sub-derivative

Example ($\ell = 3, \delta = 1$)

$$\mathcal{R}(\mathbf{x}) \xrightarrow{\text{error at } i_2} \mathcal{R}(\mathbf{x})' = (1, 1, 2, 2, 2, 1, 1, 0)$$

$$\bullet \sum_k \Delta^0(\mathbf{x})'_k = -\sum_k \Delta^1(\mathbf{x})'_k = 1 \implies i_2 \bmod \ell = 1 \implies i_2 \in \{1, 4, 7\}.$$

Reconstruction with Noisy Sub-derivative

Example ($\ell = 3, \delta = 1$)

$$\mathcal{R}(\mathbf{x}) \xrightarrow{\text{error at } i_2} \mathcal{R}(\mathbf{x})' = (1, 1, 2, 2, 2, 1, 1, 0)$$

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- Left to right reconstruction:

- ▶ $x'_1 = 1.$
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- ▶ $-x'_4 = 0$ ✗

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Reconstruction with Noisy Sub-derivative

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• Left to right reconstruction:

- ▶ $x'_1 = 1.$
- ▶ $x'_4 - x'_1 = 0 \implies x'_4 = 1$
- ▶ $-x'_4 = 0 \text{ ✗}$

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• Left to right reconstruction:

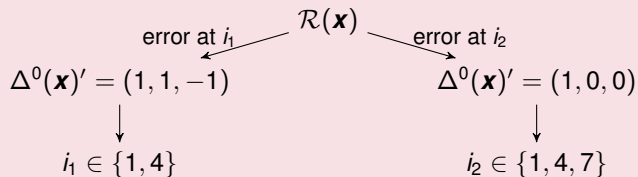
- ▶ $x'_1 = 1.$
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- ▶ $-x'_4 = 0 \text{ ✗}$

• Right to left reconstruction:

- ▶ $x'_4 = 0.$
- ▶ $x'_4 - x'_1 = 0 \implies x'_1 = 0$
- ▶ $x'_1 = 1 \text{ ✗}$

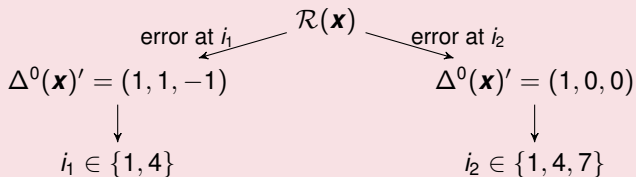
Reconstruction with Noisy Sub-derivative

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Reconstruction with Noisy Sub-derivative

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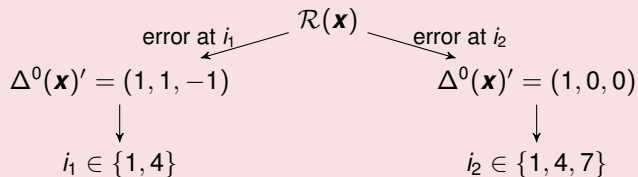


$$\Delta^0(\mathbf{x}) = (\quad 1 \quad \quad 0 \quad \quad 0 \quad \quad 0 \quad \quad \dots \quad \quad 0 \quad \quad -1)$$

$$x_1 = 1 \quad x_4 = 1 \quad x_7 = 1 \quad x_{11} = 1 \quad \dots \quad x_{\lfloor \frac{n-1}{3} \rfloor 3+1} = 1$$

Reconstruction with Noisy Sub-derivative

Example ($\ell = 3, \delta = 1$)



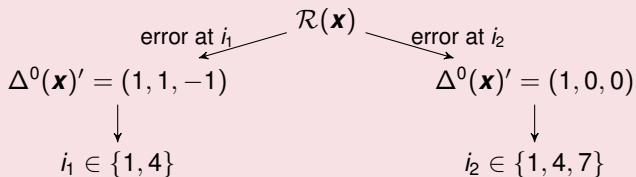
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$$x'_1 = 1 \quad x'_4 = 1 \quad x'_7 = 0 \quad x'_{11} = 0 \quad \dots \quad x'_{\lfloor \frac{n-1}{3} \rfloor 3+1} = 0$$

Reconstruction with Noisy Sub-derivative

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$$x'_1 = 1 \quad x'_4 = 1 \quad x'_7 = 0 \quad x'_{11} = 0 \quad \dots \quad x'_{\lfloor \frac{n-1}{3} \rfloor 3+1} = 0$$

- Run of 0s delays error detection \rightarrow **restrict run-length of 0s** in all $\Delta^\alpha(\mathbf{x})$!

Error-correcting Code

- restrict run-length of 0s in all read sub-derivatives,

Construction

$$\mathcal{C}(n, \ell) = \{\mathbf{x} \in \Sigma_2^n : \Delta(\mathbf{x}) \bmod 2 \in RLL(\log 2(n + \ell)),$$

where $\Delta(\mathbf{x}) = \Delta^0(\mathbf{x}) \circ \Delta^1(\mathbf{x}) \circ \dots \circ \Delta^{\ell-1}(\mathbf{x})$

C. Schoeny *et al.*, "Codes correcting a burst of deletions or insertions," *IEEE Trans. Inf. Theory*, vol. 63, no. 4, pp. 1971–1985, Apr. 2017

Error-correcting Code

- restrict run-length of 0s in all read sub-derivatives,
- Correct 1 substitution in length- a window with folded Hamming code $\mathcal{H}(n, a)$:

$$\underbrace{[\mathbf{H}_a \quad \mathbf{H}_a \quad \cdots \quad \mathbf{H}_a]}_{\frac{n}{2^a - 1} \text{ times}}$$

where \mathbf{H}_a corresponds to Hamming code of order a .

Construction

$$\mathcal{C}(n, \ell) = \{\mathbf{x} \in \Sigma_2^n : \Delta(\mathbf{x}) \bmod 2 \in RLL(\log 2(n + \ell)), \\ \mathcal{R}^\pi(\mathbf{x}) \bmod 2 \in \mathcal{H}(n + \ell - 1, \log \log 8(n + \ell) + 1)\},$$

where $\Delta(\mathbf{x}) = \Delta^0(\mathbf{x}) \circ \Delta^1(\mathbf{x}) \circ \cdots \circ \Delta^{\ell-1}(\mathbf{x})$ and $\mathcal{R}^\pi(\mathbf{x})$ is a permutation of $\mathcal{R}(\mathbf{x})$.

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→ Requires $\log \log n + o(1)$ redundant bits (optimal up to a constant).

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Introduction

Channel Model

Minimum Redundancy

Error-correcting Code

Conclusion

Summary

Results

For a simplified model of nanopore sequencing,

- Established that $\log \log n$ redundancy needed to correct 1 substitution
- Proposed an error-correcting construction that is optimal up to a constant.

Future work

- Multiple substitutions
- Deletions
- Extension to non-binary, $\delta > 1$.

Thank you!

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Correctable Substitutions

Example ($\ell = 3, \delta = 1$)

We receive $\mathcal{R}(\mathbf{x})' = (0, 1, 3, 2, 2, 2, 1, 0, 0)$.

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We receive $\mathcal{R}(\mathbf{x})' = (0, 1, 3, 2, 2, 2, 1, 0, 0)$.

- $|\mathcal{R}(\mathbf{x})'_3 - \mathcal{R}(\mathbf{x})'_2| \not\leq 1 \rightarrow$ Error pos. 2 or 3.

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 - ▶ $\mathcal{R}(\mathbf{x})_1 = \text{wt}(x_1) = 0 \rightarrow x_1 = 0$

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 - ▶ $\mathcal{R}(\mathbf{x})_3 = \text{wt}(\mathbf{x}_1^3) = 3 \rightarrow x_3 = 2 \notin \{0, 1\} \rightarrow$ Error pos. 3.

Correctable Substitutions

Example ($\ell = 3, \delta = 1$)

We receive $\mathcal{R}(\mathbf{x})' = (0, 1, \textcircled{*}, 2, 2, 2, 1, 0, 0)$.

- $|\mathcal{R}(\mathbf{x})'_3 - \mathcal{R}(\mathbf{x})'_2| \not\leq 1 \rightarrow$ Error pos. 2 or 3.
- Left-to-right reconstruction:
 - ▶ $\mathcal{R}(\mathbf{x})_1 = \text{wt}(x_1) = 0 \rightarrow x_1 = 0$
 - ▶ $\mathcal{R}(\mathbf{x})_2 = \text{wt}(\mathbf{x}_1^2) = 1 \rightarrow x_2 = 1$
 - ▶ $\mathcal{R}(\mathbf{x})_3 = \text{wt}(\mathbf{x}_1^3) = 3 \rightarrow x_3 = 2 \notin \{0, 1\} \rightarrow$ Error pos. 3.
- Consider erasure.

Correctable Substitutions

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- Left-to-right reconstruction:
 - ▶ $\mathcal{R}(\mathbf{x})_1 = \text{wt}(x_1) = 0 \rightarrow x_1 = 0$
 - ▶ $\mathcal{R}(\mathbf{x})_2 = \text{wt}(\mathbf{x}_1^2) = 1 \rightarrow x_2 = 1$
 - ▶ $\mathcal{R}(\mathbf{x})_3 = \text{wt}(\mathbf{x}_1^3) = 3 \rightarrow x_3 = 2 \notin \{0, 1\} \rightarrow$ Error pos. 3.
- Consider erasure.
- $\sum_i \mathcal{R}(\mathbf{x})_i \bmod \ell = 0$

Correctable Substitutions

Example ($\ell = 3, \delta = 1$)

We receive $\mathcal{R}(\mathbf{x})' = (0, 1, 1, 2, 2, 2, 1, 0, 0)$.

- $|\mathcal{R}(\mathbf{x})'_3 - \mathcal{R}(\mathbf{x})'_2| \not\leq 1 \rightarrow$ Error pos. 2 or 3.
- Left-to-right reconstruction:
 - ▶ $\mathcal{R}(\mathbf{x})_1 = \text{wt}(x_1) = 0 \rightarrow x_1 = 0$
 - ▶ $\mathcal{R}(\mathbf{x})_2 = \text{wt}(\mathbf{x}_1^2) = 1 \rightarrow x_2 = 1$
 - ▶ $\mathcal{R}(\mathbf{x})_3 = \text{wt}(\mathbf{x}_1^3) = 3 \rightarrow x_3 = 2 \notin \{0, 1\} \rightarrow$ Error pos. 3.
- Consider erasure.
- $\sum_i \mathcal{R}(\mathbf{x})_i \bmod \ell = 0 \rightarrow \mathcal{R}(\mathbf{x})_3 = 1.$

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 - ▶ $\mathcal{R}(\mathbf{x})_3 = \text{wt}(\mathbf{x}_1^3) = 3 \rightarrow x_3 = 2 \notin \{0, 1\} \rightarrow$ Error pos. 3.
- Consider erasure.
- $\sum_i \mathcal{R}(\mathbf{x})_i \bmod \ell = 0 \rightarrow \mathcal{R}(\mathbf{x})_3 = 1$.

Takeaway

If error magnitude > 1 , can correct immediately.
 \rightarrow Sufficient to consider ± 1 errors!