Stanford

Overcoming high nanopore basecaller error rates for DNA storage via basecaller-decoder integration and convolutional codes

Shubham Chandak Stanford University ICASSP 2020

Team and funding











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SemiSynBio: Highly scalable random access DNA data storage with nanopore-based reading

Beckman Center Innovative Technology Seed Grant Scalable Long-Term DNA Storage with Error Correction and Random-Access Retrieval



National Institutes of Health

Motivation



40,000 x 5 TByte HDDs 40 tons

10s of years



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10s of years



DNA 1 gram

1,000s of years



40,000 x 5 TByte HDDs 40 tons

10s of years



1,000s of years

DNA storage setup

Building block: synthesis

• Ability to "write/synthesize" artificial DNA (sequence of {A,C,G,T})



Current ability: short ssDNA oligos (~150nt) at scale

DNA Synthesis is not perfect: Usually has ~1% insertion/Deletion error

Building block: sequencing

• Nanopore sequencing: portable, real time



https://directorsblog.nih.gov/2018/02/06/sequencing-human-genome-with-pocket-sized-nanopore-device/

Typical DNA Storage System



Challenges

- High basecall error rates for nanopore sequencing
 - 5-10% edit distance
 - Predominantly insertion and deletion errors
- Lack of good error correction codes for this setting

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- High basecall error rates for nanopore sequencing
 - 5-10% edit distance
 - Predominantly insertion and deletion errors
- Lack of good error correction codes for this setting
- Most previous works rely on consensus over multiple reads high reading cost
 - Sequence the input lot of times (~30-40x)
 - Cluster by *index*, and perform "averaging" to reduce the error

Previous Works



[2] L. Organick *et al.*, "Random access in large-scale DNA data storage," *Nature biotechnology*, vol. 36, no. 3, p. 242, 2018.

[3] Randolph Lopez et al., "DNA assembly for nanopore data storage readout," Nature communications, vol. 10, no. 1, pp. 2933, 2019.

Methods

Nanopore Physics



Nanopore Sequencing Model



Nanopore sequencing channel

- Memory (inter-symbol interference)
- Base skips
- Fading
- Random symbol duration
- Noise



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VERY HARD TO MODEL AND ANALYZE FAITHFULLY

Source: "Models and Information-Theoretic Bounds for Nanopore Sequencing", Wei Mao et al., IEEE Trans. Inf. Theory 2017

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VERY HARD TO MODEL AND ANALYZE FAITHFULLY

COMBINE STRENGTHS OF MACHINE LEARNING & CODING THEORY!

Source: "Models and Information-Theoretic Bounds for Nanopore Sequencing", Wei Mao et al., IEEE Trans. Inf. Theory 2017



Key idea



Using Flappie basecaller (Oxford Nanopore)

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Key idea



Convolutional Codes as the Inner Code



Basecaller-decoder integration



NN-modeling based transition probabilities

Overall Inner Code design



Experiments and results

Experiments

- Data: 11KB of data: The Gettysburg Address, UN Declaration, "I have a Dream" Speech, poem collections, ...
- Final Error Correction Code Design:
 - Reed Solomon outer code: 30% redundancy (default)
 - Pretrained Model from the ONT Flappie Basecaller
- Synthesis: Data Synthesized using CustomArray synthesis, into oligos of length ~165
- Experiments:
 - Rate of convolution code: r = 1/2, 3/4, 5/6
 - Memory: m = 8,11,14
 - List Size: 4, 8

Results



[6] L. Organick *et al.*, "Random access in large-scale DNA data storage," *Nature biotechnology*, vol. 36, no. 3, p. 242, 2018.

[22] Randolph Lopez et al., "DNA assembly for nanopore data storage readout," Nature communications, vol. 10, no. 1, pp. 2933, 2019.

Conclusions and future work

- Novel error-correction mechanism for nanopore sequencing based DNA storage
 - Use "soft-information" from raw signal to improve decoding
 - Use neural net in basecaller to distil information from "hard-to-model" raw signal
 - Use convolutional codes that align nicely with sequential nanopore model
- Requires 3x fewer reads for decoding than previous works

Conclusions and future work

- Novel error-correction mechanism for nanopore sequencing based DNA storage
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- Future work:
 - Optimization of convolutional code and CRC parameters
 - Finetuning of neural network model and use of improved basecallers
 - Application to other novel synthesis methodologies

Thank You!

Code and data available at

https://github.com/shubhamchandak94/nanopore_dna_storage