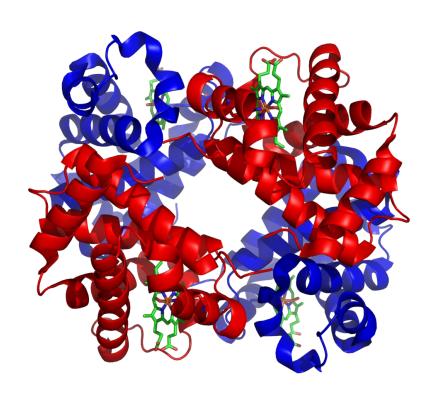
Evaluating Protein Transfer Learning with TAPE

Roshan Rao*, Nicholas Bhattacharya*, Neil Thomas*,
Yan Duan, Xi Chen,
John Canny, Pieter Abbeel, Yun S. Song

Why care about proteins?

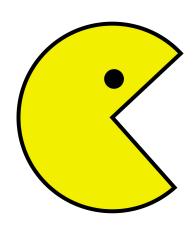
Hemoglobin

Antibodies





Tired of eating plastic? Call 1-800-PROTEIN



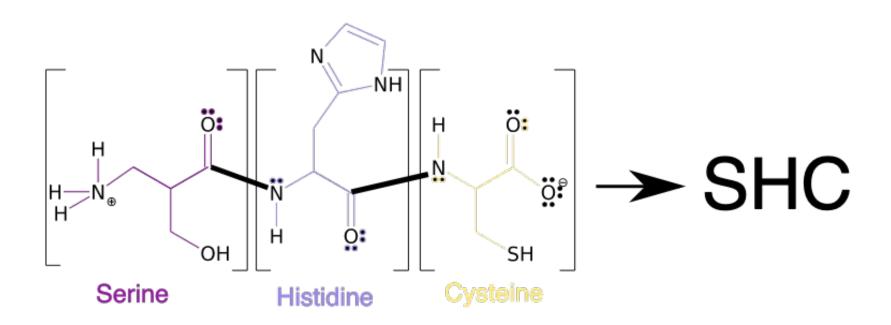




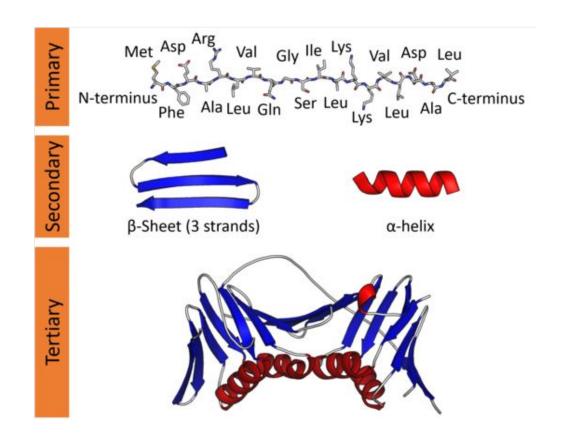
What is a protein?

A 3 slide crash course

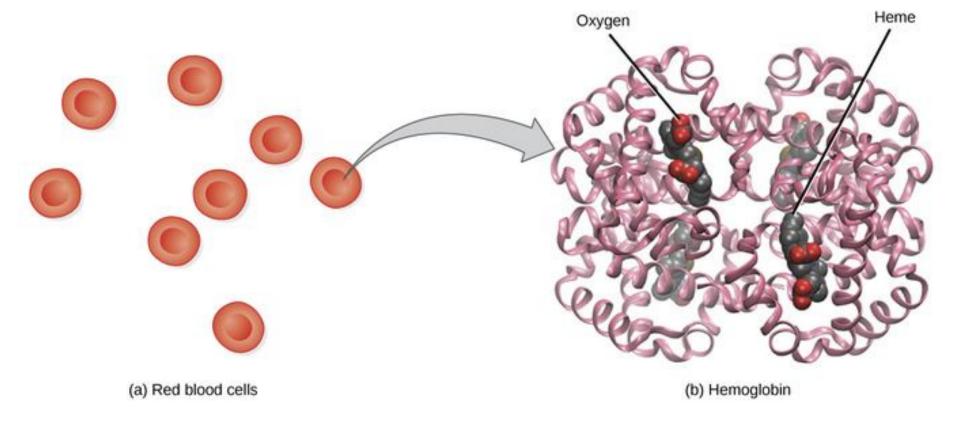
Sequence



Structure



Function



How do we find new sequences?

Collecting unlabeled data easy



1. Put on protective equipment

2. Collect dirt

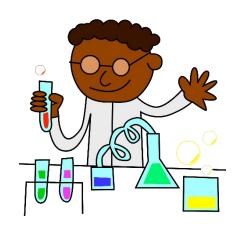


4. Lots of Sequences (Genes)



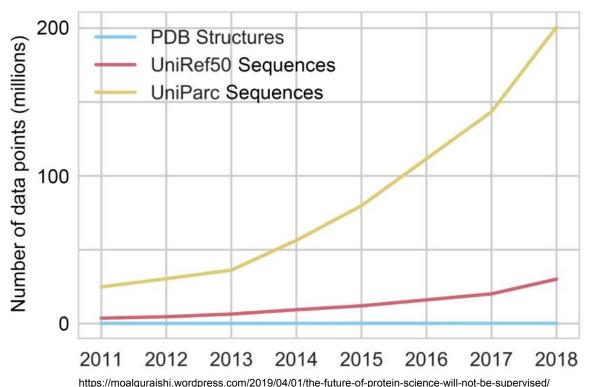
3. Throw it in the sequencer

How do we determine new structures?



Collecting labeled data hard

We cannot keep up with the sequence explosion



200 Million sequences

150,000 structures

Can we take what we've learned from detailed experimental characterization and generalize to unseen portions of sequence space?

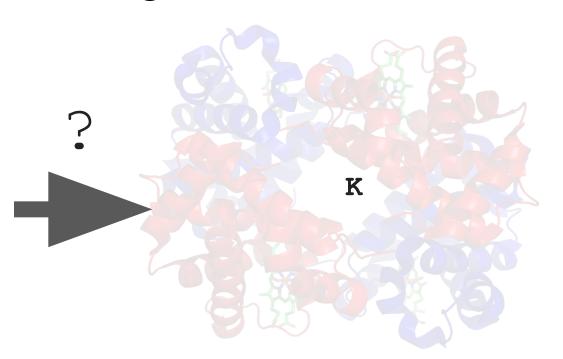
Pretrained models such as BERT make efficient use of labeled data





"Language Modeling" for Proteins

MVLSPADKTNV[?]AAWGKVG
AHAGEYGAEALE...



Which one is better?

- Different downstream tasks
- Different pretraining corpuses
 - o 20 million sequences
 - o 200 million sequences
- Different compute budgets
 - 1 GPU
 - o 128 GPUs

LEARNING PROTEIN SEQUENCE EMBEDDINGS USING INFORMATION FROM STRUCTURE

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nature **methods**

Article | Published: 21 October 2019

Unified rational protein engineering with sequence-based deep representation learning

Ethan C. Alley, Grigory Khimulya, Surojit Biswas, Mohammed AlQuraishi & George M. Church

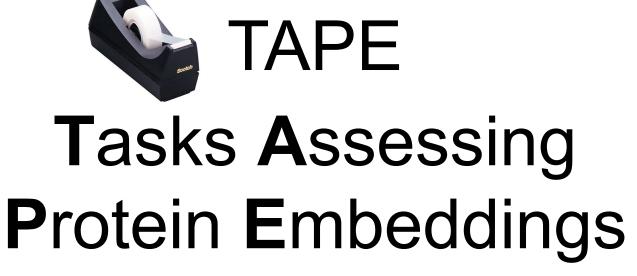
Nature Methods (2019) | Download Citation ±

142 Altmetric | Metrics >>

BIOLOGICAL STRUCTURE AND FUNCTION EMERGE FROM SCALING UNSUPERVISED LEARNING TO 250 MILLION PROTEIN SEQUENCES

Alexander Rives *† \$\frac{1}{5}\$ Siddharth Goyal *\frac{5}{5}\$ Joshua Meier *\frac{5}{5}\$ Demi Guo *\frac{5}{5}\$

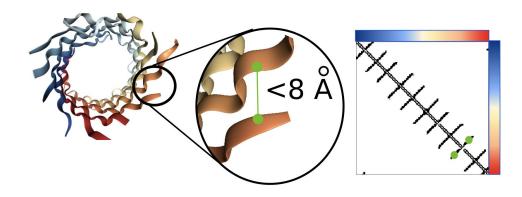
Myle Ott \frac{5}{5}\$ C. Lawrence Zitnick \frac{5}{5}\$ Jerry Ma †\frac{5}{5}\$ Rob Fergus †\frac{5}{5}\$



- Fixed 5 downstream tasks from different domains of protein biology testing meaningful generalization
- Pretrained 5 different models with:
 - Fixed corpus: 30 million sequences of protein domains
 - Fixed budget: 1 week on 4 NVIDIA V100s

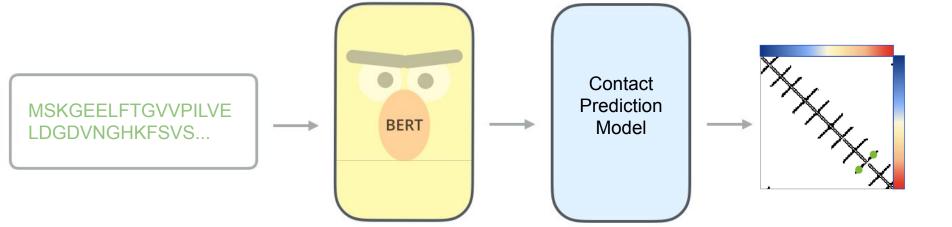


Contact Prediction

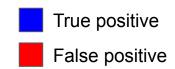


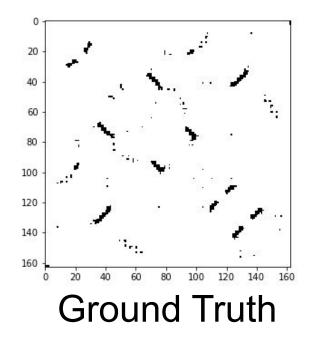


Input Features Output Prediction

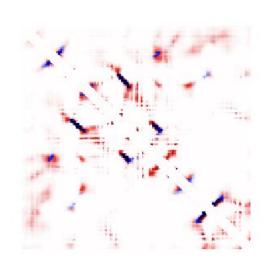


Pretraining Helps



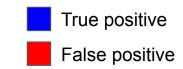


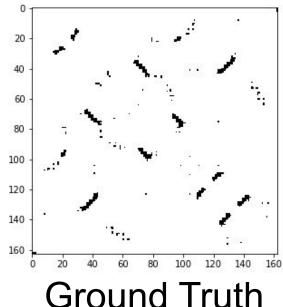
No Pretraining (LSTM)



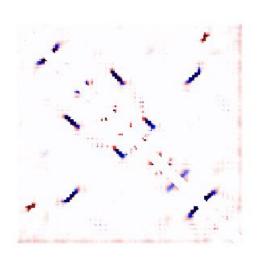
Pretrained (LSTM)

Unused signal remains!

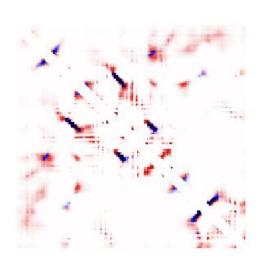




Ground Truth



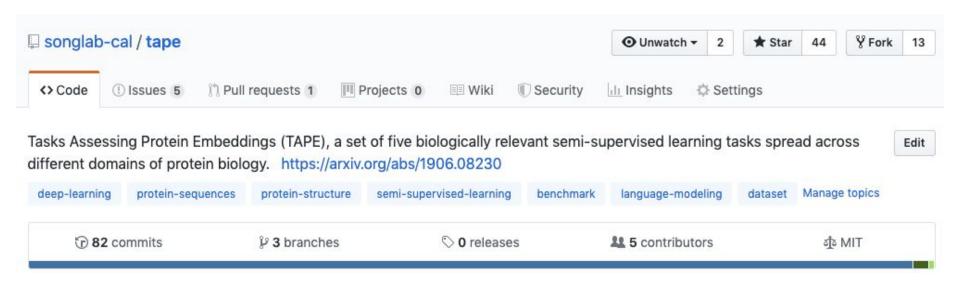
Non-neural features



Pretrained (LSTM)

Data/code for benchmark available

https://github.com/songlab-cal/tape



Come chat more at our poster! (#79)



of Health















Roshan Rao, Nick Bhattacharya, Neil Thomas, Rocky Duan, Peter Chen, John Canny, Pieter Abbeel, Yun S. Song



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