

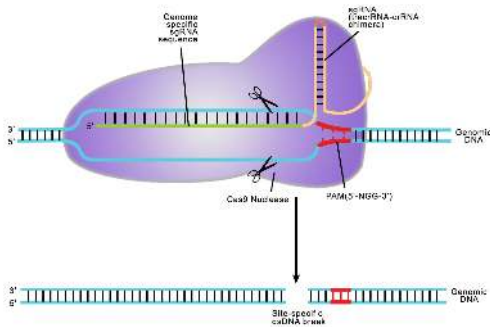


PAVOOC - Prediction and visualization of on- and off-targets for CRISPR

Moritz Schäfer | Technische Universität Berlin & Bayer Pharma | Master thesis



Background



- Knockout experiments used in drug target validation
- Sequence (partially) determines efficacy



Problem

- Guide prediction scores still vary in performance





Problem

- Guide prediction scores still vary in performance
- A lot of manual labor for guide selection

19	POLR2L	28,5	28	-3,0075211818262
20	SPC24	29,5	42	-2,83403119872683
21	THAP1	31,5	27	-3,01155447143932
22	CDC123	32,5	44	-2,81699463079959
23	WDR74	37	68	-2,58033320817648

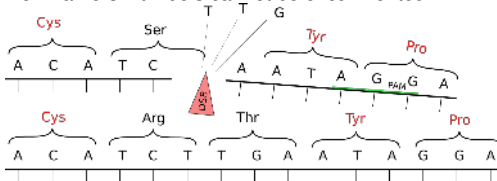


Problem

- Guide prediction scores still vary in performance
- A lot of manual labor for guide selection

19	POLR2L	28,5	28	-3,0075211818262
20	SPC24	29,5	42	-2,83403119872683
21	THAP1	31,5	27	-3,01155447143932
22	CDC123	32,5	44	-2,81699463079959
23	WDR74	37	68	-2,58033320817648

- Non-frame-shift indels cannot be circumvented



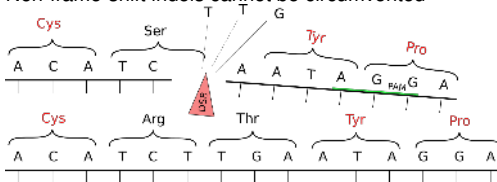


Problem

- Guide prediction scores still vary in performance
- A lot of manual labor for guide selection

19	POLR2L	28,5	28	-3,0075211818262
20	SPC24	29,5	42	-2,83403119872683
21	THAP1	31,5	27	-3,01155447143932
22	CDC123	32,5	44	-2,81699463079959
23	WDR74	37	68	-2,58033320817648

- Non-frame-shift indels cannot be circumvented



- Cancer cellline data affects certain guides

AGCATCGTAAGTGAATTACGG
- + PAM

DMS153 lung cancer cellline SNP



Solution

- Cutting-edge guide efficacy scoring
- All-in-one guide design tool
- Web-based
- Functional domain-aware
- Incorporate cancer cellline data





Guide efficacy prediction – Dataset

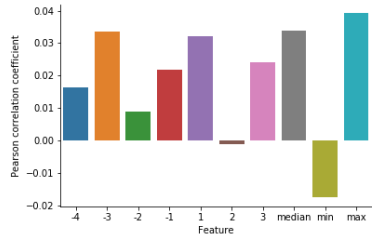
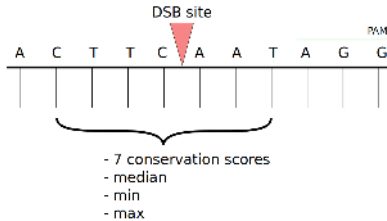
Guide	Measured efficacy
GTAGGGGTCCGTACTCAGCAAGG	0.86
ACACTGCCGAGCGATGAGGATGG	0.42
AAGGTGAAGGAGGATGCGGCGGG	0.53
GAAAAGATAGGTCACTGACCCGG	0.12
GCAAGTCACTGAGTGCAGAACGG	0.73
GCATTGGTAAGCGCACAGGAAGG	0.70
AAGACTGGCGCATGGTCCACTGG	0.57
...	...

- 5310 data rows
- Efficacy relates to cell proliferation after CRISPR application

"Optimized sgRNA design to maximize activity and minimize off-target effects of CRISPR-Cas9", 2016, John G. Doench et al.



Conservation features

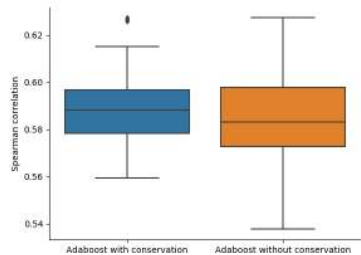
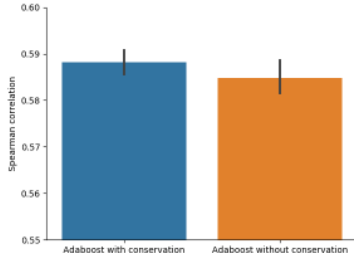


$$p_{max} = 0.0043$$



Conservation feature results

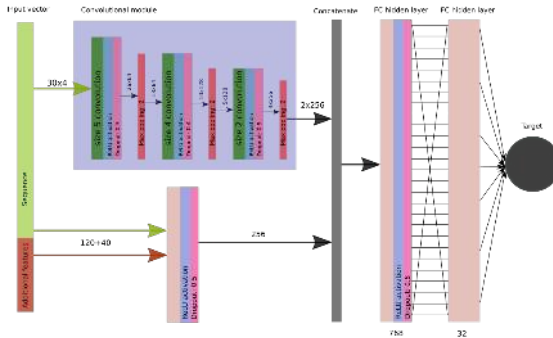
Comparison of 100 runs on adaboost





Deep Learning architecture

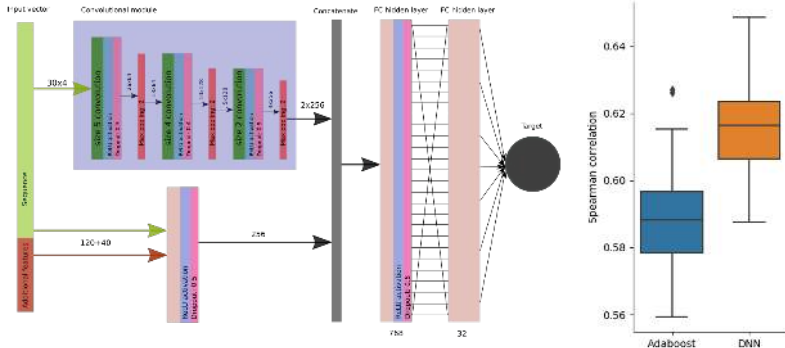
"Predicting the sequence specificities of DNA- and RNA-binding proteins by deep learning", 2015, B. Alipanahi et al.





Deep Learning architecture

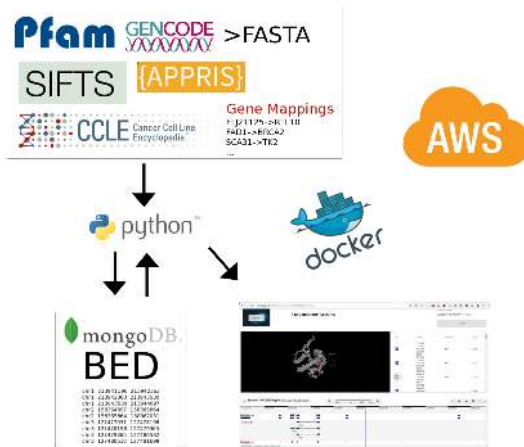
"Predicting the sequence specificities of DNA-and RNA-binding proteins by deep learning", 2015, B. Alipanahi et al.



Conclusion: Deep Learning can improve guide efficacy prediction



Architecture





Application overview

