

2 **Estimating the genetic load of natural protein coding**
3 **sequences using a phylogenetic framework.**

4 **Abstract**

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Abstract

Protein production is a very costly process every cell performs, resulting in selection for proteins that can perform their function most efficiently. The efficacy of selection is limited by the effective population size N_e , leading to genetic load via the introduction of mutations. As all proteins have to face this selection-mutation-drift barrier, we expect to find proteins near a fitness peak, but never at the peak. Here we assess the efficacy of selection on individual proteins and quantify the genetic load by phylogenetic inference of the optimal amino acid at each site using SelAC. We demonstrate the assessment of the genetic load proteins impose for TEM in *E. coli* and for cytochrome B in whales. We quantify the genetic load for 49 TEM sequences and 12 cytochrome B sequences. We compare the inferred optimal TEM amino acid sequence to fitness estimates from deep mutation scanning experiments. We find that the observed TEM sequences have a 3 to 20 fold increased genetic load when compared to the DMS estimates instead of the SelAC inference. Furthermore, we find that the DMS inference only shows 49 % sequence agreement with the consensus sequence of the observed alignment. We also observe a higher genetic load in CytB than in TEM, which is to be expected given the difference in N_e between whales and *E. coli*.

Introduction

- Genetic load is a measure of distance between the average genotype's fitness and the genotype with the highest fitness.
 - The genotype with the highest fitness is assessed based on the set of observed genotypes.
 - Mutation constantly introduces new, potentially deleterious mutations increasing the genetic load of a population.
 - Genetic drift limits the efficacy of natural selection.

– Therefore, the optimal genotype is likely not among the observed genotypes.

– To remedy this, experimental procedures like deep mutation scanning can be employed to assess the fitness of genotypes.

- Deep mutation scanning (DMS) requires a library of mutants for which the fitness should be assessed.

– This limits the application of DMS experiments to organisms that can be manipulated under laboratory conditions, and have a sufficiently short generation time.

– It also requires that artificial selection can be applied.

– This limits DMS experiments even further to proteins for which we can assume they respond to a singular stress factor.

– While it is safe to ignore effects of mutation, the low population size does severely limit the efficacy of selection.

– It is therefore required to apply extremely strong selection pressure.

- In this study, we assess how well DMS experiments can explain genetic load produced by natural evolution.

– It has previously been demonstrated that incorporation of DMS experiments into phylogenetic approaches improve model fit when compared to classical approaches like GY94.

– However, model adequacy has not been assessed.

– First we show that the models fits achieved by the incorporation of DMS experiments can be improved upon using a novel phylogenetic framework, SelAC.

– We compare the genetic load of natural TEM variants according to DMS and SelAC and show that DMS predicts an increased genetic load.

– We find that we would not expect to observe the natural TEM variants when simulating under the DMS inferred fitness landscape.

- Having shown that SelAC provides more adequate inference of genetic load we further demonstrate its generality by assessing the genetic load of cytochrome B in whales.

Results

- We compared how well deep mutation scanning experiments can explain observed sequences and compared it to SelAC.

– We utilized DMS results from Stiffler et al. for β -lactamase in *E. coli*.

– We utilized phydms a tool designed to incorporate fitness estimates obtained from the DMS experiment into a phylogenetic framework.

– Model selection reveals that SelAC outperforms the the DMS experiment and better explains the observed sequences.

– Furthermore, we estimated the expected sequences given the fitness landscape described by DMS and SelAC.

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– Simulated sequences evolve to about 60 % sequence similarity.

– We would also expect to observe less genetic load as sequences had time to evolve, assuming a reasonable N_e of 1,000 or greater.

- We estimated the functionally optimal amino acid at each site from the observed sequence variation using SelAC.

– The observed TEM alignment shows a high percentage of homogeneous sites.

* 68% of sites had only one codon present.

* 75% of sites encoded the same amino acid.

84 – The observed CytB alignment shows a more codon heterogeneity but a similar
85 homogeneity in amino acids.

86 * 22% of sites had only one codon present.

87 * 78% of sites encoded the same amino acid.

88 – We find that the predicted optimal amino acid sequence has high agreement with
89 the observed consensus sequence of the alignment (TEM: 99%, CytB: 95%).

90 – In contrast, the experimentally obtained sequence estimate only has an agreement
91 of 49% with the observed TEM consensus sequence.

92 • We assessed the genetic load of the observed sequences.

93 – We find that the genetic load of TEM differs greatly depending on the optimal
94 amino acid sequence assumed.

95 * The genetic load of the observed sequences increases 3 – 20 fold when using
96 the experimentally inferred optimal sequence compared to the SelAC inferred
97 optimal sequence.

98 * Besides the great variation that arises from the usage of different optimal
99 amino acid sequences, we also find variation within each optimal amino acid
100 sequence.

101 * E.g. sN_e varies between ~ 0 to ~ -10 for the SelAC optimal sequence and
102 between ~ -20 to ~ -27 for the optimal sequence obtained from the DMS
103 experiment.

104 – We lack the ability to compare our estimates of genetic load for CytB as DMS
105 experiment can not be performed on whales.

106 – We find a higher genetic load and greater variation in sN_e for the CytB (not taken
107 into account: sequences differ in length).

108 * sN_e varies between ~ -10 to ~ -35 .

- 109 • We are able to map variation in selection along the sequence and determine sites with

110 higher contribution to genetic load.

 - 111 – Increases in genetic load appear to be locally confined to a few regions among the

112 TEM alignment but do not appear to be associated with any particular structural

113 features.
 - 114 – In contrast, CytB shows variation of genetic load across its whole sequence with a

115 particularly strong increase in genetic load within the 5th transmembrane helix.
- 116 • Previous work highlighted the advantages of DMS experiments for phylogenetic infer-

117 ences.
- 118 • However, our estimates of genetic load of observed TEM sequences show that natural

119 sequences would actually represent a large genetic load.

 - 120 – The SelAC estimated optimal amino acid sequence outperformed the consensus

121 sequence and the experimentally sequence explaining the data.
 - 122 – A second model selection was performed using phydms as an independent com-

123 parison.

 - 124 * Model selection revealed that the main advantage of the DMS experiment

125 comes from the fact that the input alignment is not needed to estimate amino

126 acid preferences.
 - 127 * While the experimentally inferred optimal sequence does a worse job explain-

128 ing the observed sequences, model selection reveals that the improvement in

129 likelihood does not justify the increased number of parameters required to

130 run phydms with the SelAC or the consensus amino acid preferences.

Discussion

- We demonstrate the inference of site specific selection from protein coding sequence data using phylogenetics.
 - We find that the fitness landscape estimated by SelAC better explains the observed sequences than DMS.
 - Simulations show that the observed sequences would not arise under the imposed selection during the DMS experiments.
 - While DMS allows for the inference of properties such as substitutions conferring antibiotic resistance, it does not allow to explain natural sequence variation.
 - In addition, as researchers show more and more interests in non-model organisms, the limitation to proteins and organisms that can be manipulated under laboratory conditions limits its uses across the tree of life.
- We estimate the genetic load of natural occurring proteins relative to an inferred optimal amino acid sequence.
- The optimal amino acid at each site was inferred from the observed proteins and their phylogenetic relationship.
- In both cases, TEM and CytB, we find high agreement between the consensus sequence inferred by ignoring the phylogenetic relationship and the optimal sequence inferred using SelAC (TEM: 99%, CytB: 95%).
 - The strong agreement between consensus sequence and estimated optimal sequence for both proteins can be seen as an indication that the phylogenetic relationship does not play a large role in the examined cases.
 - However, such an assumption should not be made a priori.

- The similarity between consensus and predicted optimal sequence could be because the proteins are under stabilizing selection like the model assumes, because rate of shifts in the optimal amino acid sequence is low, or because not enough time has passed for shifts to occur, despite diversifying selection.
- The used alignments contain a high amount of homogeneous sites (TEM: 75%, CytB: 78%), thus these sites do not allow for the inferred optimal amino acid to deviate from the observed consensus.
- In contrast, the experimentally inferred optimal amino acid sequence for TEM only has 49% agreement with the observed consensus.

 - Assuming that this inferred sequence is free of any bias introduced by the experimental conditions, we could only come to the conclusion that the observed TEM sequences show either strong mal-adaptation or did not have enough time to evolve towards the optimal sequence.
 - However, *E. coli* has a large effective population size, estimates are on the order of 10^8 to 10^9 (Ochman and Wilson 1987, Hartl et al 1994).
 - The large N_e would allow *E. coli* to effectively "explore" the sequence space.
 - On the other hand, each mutation in the library used for the DMS experiments starts of with only a few copies, potentially biasing the results due to strong genetic drift.
- The genetic load of the observed sequences was inferred relative to the optimal amino acid sequence estimated by SelAC.

 - Both, CytB and TEM show variation in the genetic load represented by each observed sequence, CytB represents a higher genetic load than TEM.
 - Most TEM sequences show a small genetic load, likely due to the high selection pressure on TEM due to its usage in chemical warfare between microorganisms.

- * If the experimental sequence is assumed to be most optimal, the observed TEM proteins represent a high genetic load to the organism.
 - * This would be in conflict with a large effective population size and therefore high efficacy of selection.
 - * However, while this would make fixation unlikely, it would not be impossible.
 - * In addition, the experimental sequence was inferred based on small population sizes for each genotype and artificial selection pressure.
- Genetic load varies across the sequence.
 - For both proteins, variation of sN_e across the sequence is not associated with any particular structural features but mostly with variation in the alignment.
 - However, TEM shows increased genetic load near the binding site, and the highest genetic load is found in the last beta sheet of the protein.
 - The genetic load is generally higher for CytB than for TEM, and like for TEM genetic load appears to increase around the binding sites.
 - However, for both proteins, increases in genetic load are not limited to the binding sites.
 - DMS experiments have been incorporated into phylogenetic studies to supplement information on selection on amino acids.
 - In contrast, this study shows that information on selection can be extracted from alignments of protein coding sequences.
 - To no surprise, model selection clearly favored the optimal sequence inferred by SelAC when using SelAC, however, when using this sequence in phydms we find that the inferences from SelAC still explained the data better, but the increase in parameters did not merit the increase in likelihood.
 - This highlights the limitations of DMS sequences to explain natural evolution.

Model	L	n	AIC	Δ AIC
SelAC	-1498	374	3744	0
SelAC+DMS	-1768	111	3758	14
phyDMS	-2060	105	4331	586

Table 1: L , number of model parameters n , AIC, and Δ AIC.

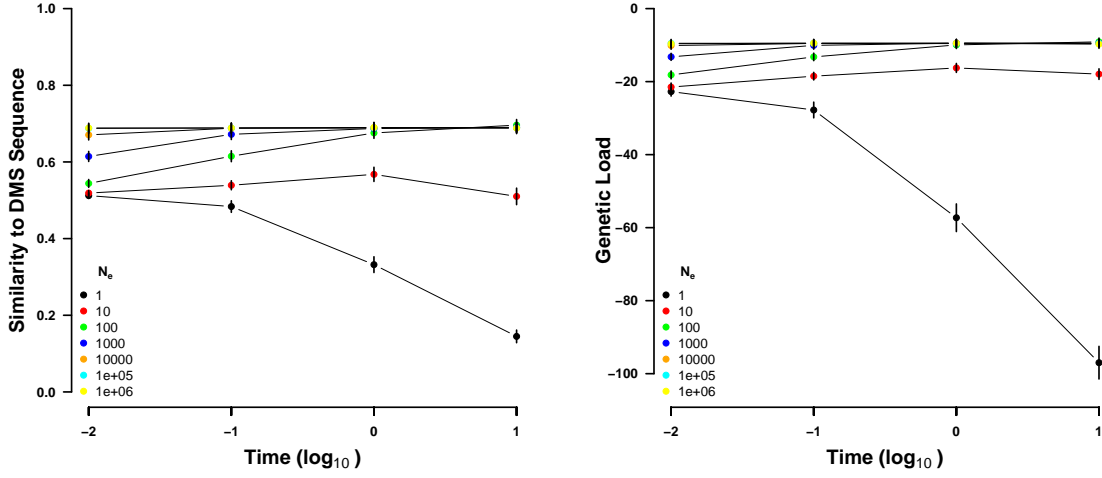


Figure 1: Sequences simulated under various values of N_e and for various times.

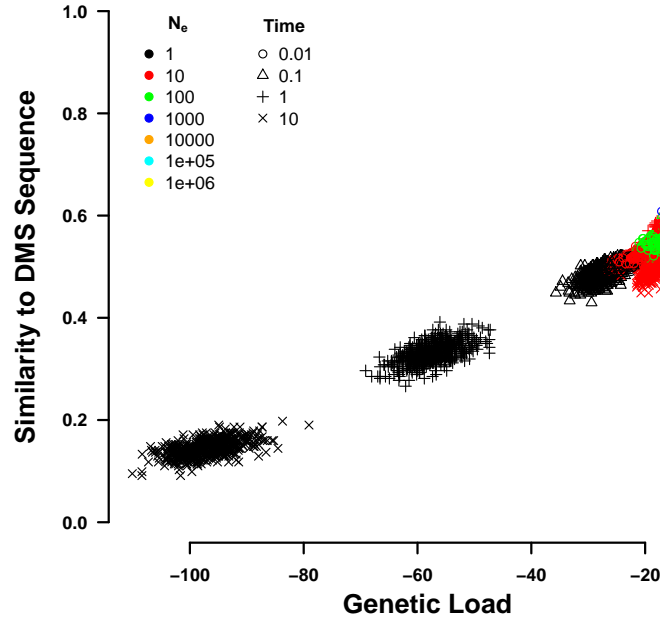


Figure 2: Sequences simulated under various values of N_e and for various times.

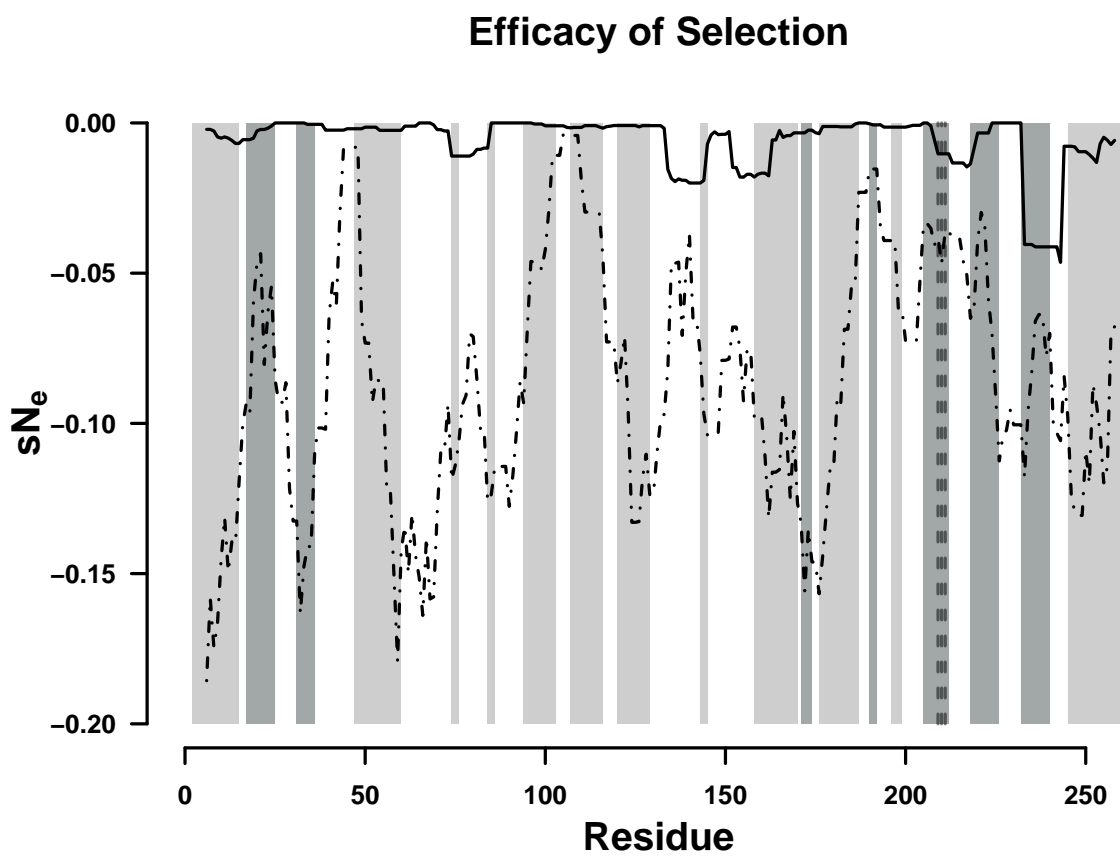


Figure 5: TEM, bars are different secondary structure elements, dashed line is DMS sN_e , solid is SelAC., horizontal lines are active/binding sites.

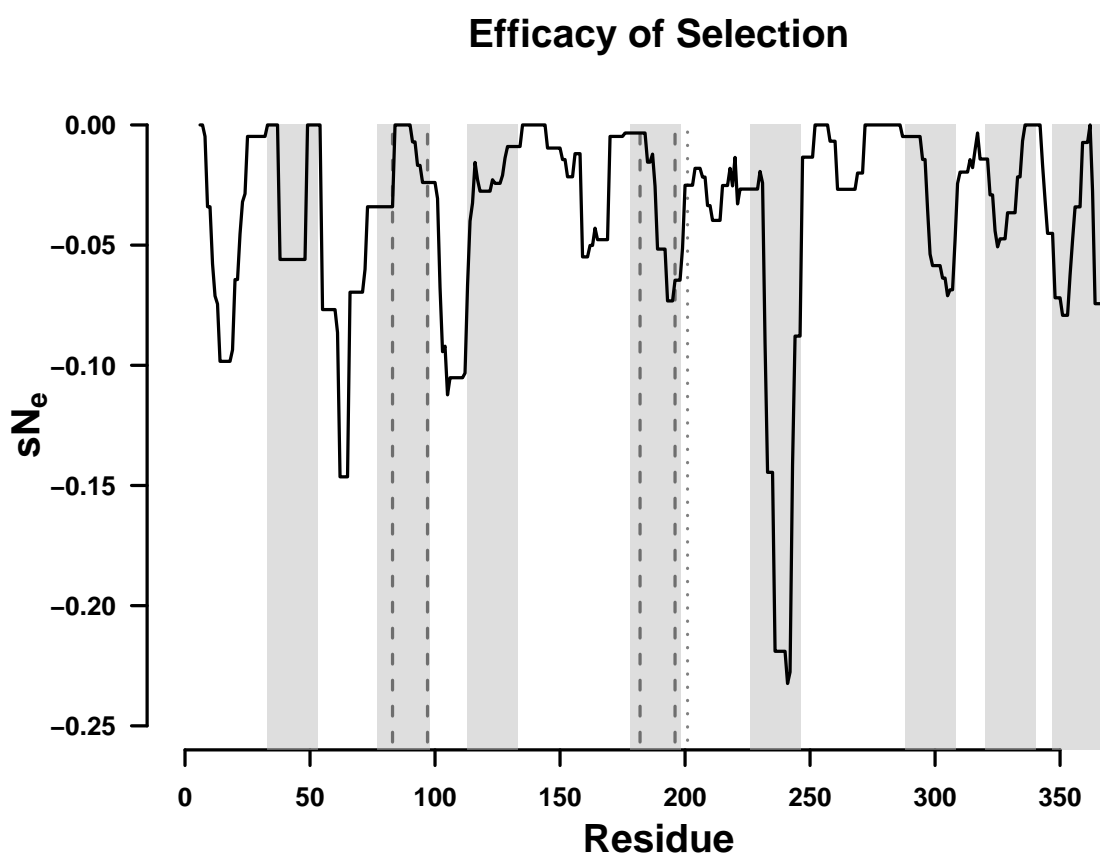


Figure 6: CytB