Active site center redesign increases protein stability preserving catalysis in thioredoxin

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Abstract
The stabilization of natural proteins is a long-standing desired goal in protein engineering. Optimizing the hydrophobicity of the protein core often results in extensive stability enhancements. However, the presence of totally or partially buried catalytic charged residues, essential for protein function, has limited the applicability of this strategy. Here, focusing on the thioredoxin, we aimed to augment protein stability by removing buried charged residues in the active site without loss of catalytic activity. To this end, we performed a charged-to-hydrophobic substitution of a buried and functional group, resulting in a significant stability increase yet abolishing catalytic activity. Then, to simulate the catalytic role of the buried ionizable group, we designed a combinatorial library of variants targeting a set of seven surface residues adjacent to the active site. Notably, more than 50% of the library variants restored, to some extent, the catalytic activity. The combination of experimental study of 2% of the library with the prediction of the whole mutational space by partial least squares regression revealed that a single point mutation at the protein surface is sufficient to fully restore the catalytic activity without thermostability cost. As a result, we engineered one of the highest thermal stabilities reported for a protein with a natural occurring fold (137°C). Further, our hyperstable variant preserves the catalytic activity both in vitro and in vivo.

KEYWORDS
genotype–phenotype mapping, partial least squares reconstruction, protein design, protein evolution, protein stability, thioredoxin

1 | INTRODUCTION
Most proteins exhibit nearly balanced free energy profiles for folded and unfolded states1 which hampers their biotechnological applications. Furthermore, as most mutations are destabilizing,2 marginal stability becomes a significant bottleneck for the laboratory evolution3,4 and computational design of proteins.5,6 In consequence, many efforts have been directed to increase protein stability.1,7–10 Notably, de novo design has been revealed as
one of the most successful strategies to achieve protein hyperstability, partly attributed to the well-packed and exclusively hydrophobic cores of its designs.\textsuperscript{11} Indeed, optimizing natural proteins’ core to maximize its buried hydrophobic surface area often results in extensive stability enhancements.\textsuperscript{12–17} However, it is frequently impossible to fully implement such a strategy in natural proteins, as buried ionizable residues are often part of active sites and, thus, linked to function.\textsuperscript{18–20} Therefore, when applying this stability strategy to natural proteins, functional buried ionizable residues typically remain untouched.

This trade-off between stability and function is exemplified by the electron transfer protein thioredoxin. The thioredoxin\textsuperscript{21} (Trx) is a small and well-characterized protein with an active site center partially exposed.\textsuperscript{22} It comprises a tryptophan followed by two vicinal cysteines spaced by a glycine-proline segment (Trp-Cys\textsubscript{32}-Gly-Pro-Cys\textsubscript{35}), enumerated according to the sequence of the \textit{E. coli}-Trx, and a conserved and buried aspartic residue (Asp\textsubscript{26}).\textsuperscript{21–23} Mutating the buried Asp\textsubscript{26} to a hydrophobic residue dramatically drops the catalytic activity\textsuperscript{23,24} yet increases the protein stability.\textsuperscript{14,21}

We aim to increase protein thermostability by replacing a buried and functional charged group with a hydrophobic residue without activity cost. It has been previously shown how the rational redesign of the surface charge distribution can modulate the \( pK_a \) and the ionization state of a buried residue.\textsuperscript{25} In this work, we go a step further and aim to simulate a buried ionizable group’s functional role, comprised in the active site center of a protein, by redesigning the surface charge distribution adjacent to the active site. Such a strategy may allow maximizing the buried hydrophobic area of proteins while preserving their catalytic capabilities.

Here, we optimized the thioredoxin’s protein core by mutating its buried and functional Asp\textsubscript{26} to Ile. Then, based on structural criteria, we selected a set of solvent-exposed positions that might balance the absence of the buried aspartic residue. After that, we constructed a combinatorial library of thioredoxin variants where the previously selected positions were mutated to lysine, glutamate, or kept the wild-type residue. We tested the thermostability and catalytic activity of 39 variants accounting for 1.8% of the total library of 2,187 variants. We used partial least squares (PLS) reconstruction to predict the whole mutational space of the library. Lastly, we selected those variants with predicted higher redox activity and experimentally validated their thermostability and activity in vitro and in vivo.

Notably, our results show that about half of the library’s variants restored the catalytic activity of the thioredoxin to some degree. Strikingly, introducing a sole lysine on the protein surface was enough to completely restore the activity of the conserved and buried aspartic residue in vitro. This redesigned active site variant was functional in vivo and displayed one of the highest thermal stabilities ever reported.\textsuperscript{26} These findings support that neutral-to-charged mutations of solvent-exposed residues can mimic the select physicochemical properties of catalytic buried ionizable groups. Furthermore, this potential might be harnessed to stabilize natural proteins by substituting functional yet destabilizing buried charged groups without compromising their catalytic activities.

## 2 | RESULTS

### 2.1 | Surface mutations simulate the catalytic role of a buried charged residue in thioredoxin in vitro

We want to increase protein stability maximizing the protein core’s hydrophobicity. An established approach consists in substituting buried charged residues with hydrophobics.\textsuperscript{11–16} However, as buried charged groups are often engaged in functional roles, this strategy risks augmenting stability at the expense of function. Therefore, first, we aim to fashion the functional role of a buried charged residue by surface mutations directed to redesign the charge surface distribution nearby. This strategy profits from the plasticity of the protein’s surface, which allows rearranging the protein, minimizing the stability effect of the mutations.\textsuperscript{27,28}

To meet this goal, we focused on thioredoxin. Thioredoxins are proteins present in all organisms that catalyze disulfide bond reductions.\textsuperscript{29} Their canonical active site center hosts the segment Trp-Cys\textsubscript{32}-Gly-Pro-Cys\textsubscript{35}. Close to the disulfide and part of the active site center, a buried carboxylic residue (Asp\textsubscript{26}) is utterly conserved among thioredoxins (Figure 1a). The reaction catalyzed by Trx is a bimolecular nucleophilic substitution that involves shuffing two electrons from the Trx to the substrate. Firstly, the hydrophobic active site residues interact with the substrate protein. Then, in the hydrophobic environment of the complex, Cys\textsubscript{32} nucleophilically attacks the substrate’s disulfide bond, resulting in an intermediate disulfide bond between Trx and the substrate. Finally, the deprotonated Cys\textsubscript{35} nucleophilically attacks the mixed disulfide bond, resulting in the oxidation of Trx and the reduction of the substrate\textsuperscript{30} (Figure 1b). The buried Asp\textsubscript{26} deprotonates and activates the Cys\textsubscript{35},\textsuperscript{24,31,32} guiding the reaction’s molecular mechanism and removing the Asp\textsubscript{26} results in a dramatic decrease in the redox capacity of Trx.\textsuperscript{23} To test whether we can mimic the contribution to the redox activity of Asp\textsubscript{26} we used as the
library starting point a modified version of an ancestral Trx from the last universal common ancestor of the Cyanobacterial, Deinococcus, and Thermus groups, called from now on trx0, that has a melting temperature (Tm) of 128°C. Trx0 will likely accept several destabilizing mutations, allowing to explore a more expansive mutational space and, therefore, increasing the chances to accommodate catalytic residues at positions different than those found in all-natural thioredoxins.

First, we measured the effect of the D26I substitution on trx0 activity. The D26I substitution reduced the redox capacity by 72% (Figure 1c), consistent with the previously described functional role of the buried aspartic residue in Escherichia coli Trx. Then, we measured the thermostability of D26I-trx0. The D26I substitution increases by 10°C the protein stability, scaling the Tm up to 138°C, in agreement with the previously reported stability effect of the D26I mutation in E. coli Trx. Thus, D26I-trx0 is a hyperstable protein with a residual capacity to catalyze the reduction of disulfide bonds.

**TABLE 1** Selected solvent-exposed positions to simulate the functional properties of the buried Asp26

| Residue    | Position | dCys32 (Å) | ASA (Å²) |
|------------|----------|------------|----------|
| Proline    | 30       | 7.0        | 0.58     |
| Methionine | 37       | 9.4        | 0.73     |
| Alanine    | 39       | 9.5        | 0.33     |
| Arginine   | 73       | 10.1       | 0.93     |
| Serine     | 74       | 7.5        | 0.58     |
| Valine     | 91       | 10.2       | 0.56     |
| Alanine    | 93       | 8.4        | 0.33     |

*Note: In the first column is annotated the amino acids found in the position shown in the second column. In the third column it is written the distance from the sulfur atom of the catalytic Cys32 to the closest atom of the amino acid at each position, and in the fourth column, the accessible surface area.*

We selected a set of residues that, upon mutation, might simulate the buried aspartic's functional role based on structural criteria: (a) We selected those residues close
to the active site center. For this purpose, we picked all residues in a radius of 11 Å centered in the sulfur atom of the catalytic Cys32. (b) Among those residues, we discarded the buried ones, that is, we only considered residues with an accessible surface area (ASA) above 0.3. (c) Lastly, we excluded exposed residues directly involved in the redox activity, such as Trp31, Asp61, Pro76, and Lys57. and those located adjacent to the sulfur atom of the Cys32—closer than 7 Å. With these criteria, seven positions were selected as promising candidates to change the surface charge distribution and simulate the catalytic role of the buried Asp26: Pro30, Met37, Ala39, Arg73, Ser74, Val91, and Ala93 (Figure 1d and Table 1). We then designed a combinatorial library of variants where each position could be occupied by a neutral, a positive, or a negative charged residue. Positions 30, 37, 39, 74, 91, and 93 could be hosted by their wild-type neutral residue or mutated either to lysine or glutamic acid. However, the wild-type residue found in position 73 is already positively charged; therefore, in the library, this position could be occupied by its wild-type residue or mutated either to glutamic acid or to methionine (since methionine is often found in this position in an alignment of thioredoxins).

We constructed the combinatorial library and transformed it into E. coli cells. We randomly selected 104 colonies, from which 43 expressed soluble thioredoxin variants. Sequencing of the soluble variants revealed 39 unique mutants. We purified these mutants and studied their redox activity in vitro using insulin as a substrate. Remarkably, more than half of them showed their redox activity in vitro using insulin as a substrate. We bootstrapped the experimental data set to assess the prediction’s uncertainty, that is, we repeated the PLS fitting using 10 different data sets by resampling the 39 initial experimental values. Full-library reconstructions of such 10 replicas are shown in Figure 2a and their mean values are shown in Figure 2b. The reconstruction displayed an unexpected scenario, as the redox activity might reach wild-type levels with very few mutations. We validated the reconstruction by selecting the best-predicted variants with one or two mutations (A39K, A39E, P30K/A39K/S74K, A39K/K73M, and A39K/P30K) and experimentally examined their redox activity in vitro (Figure 2c). Remarkably, a single point mutation alanine—

\[
\ln A = \sum_i \delta_i p_i + \sum_{i,j} \delta_{ij} p_{ij}
\]

where \(A\) is the redox activity; \(\delta_i\) is a vector of values (1, 0, 0) when the position \(i\) is occupied by a neutral residue, (0, 1, 0) when a positively charged residue occupies it, and (0, 0, 1) when a negatively charged residue occupies it; \(p_i\) is the effect of the residue in the position \(i\) on the redox activity; \(\delta_{ij} = \delta_i \cdot \delta_j\) takes different values depending on the type of residue found in position \(i\) and \(j\) (Table S1) and \(p_{ij}\) describes the coupling effect between residues at positions \(i\) and \(j\) in the redox activity. Hence, the first term of the equation describes the effect of the individual mutations on the activity and the second, the coupling effect of the mutations on the activity, considering a possible nonadditive effect. The fitting requires 231 output parameters (21 \(p_i\) parameters and 210 \(p_{ij}\) parameters). At the same time, the number of experimental values to be fitted is only 39 (values of the redox activity of the screened library variants). We addressed the fitting using partial least squares, an established algorithm to describe the relationship between protein sequence and function. This approach has been used to predict the phenotypic outcome of the presence or absence of mutations. Here, we further increase the mutational space to explore three different alternatives per position (a neutral, a positive, and a negatively charged amino acid) and consider the synergistic effect among mutations. PLS can predict a significant number of fitting parameters using very few experimental data because it first reduces the number of independent variables (\(p_i\) and \(p_{ij}\)) to fewer latent variables. These latent variables are orthogonal combinations of the original independent variables that define most of their variance and explain the correlation between the independent and dependent variables (\(A\), the redox activity). Thus, PLS uses the two data sets’ information, the dependent and independent variables, to define latent variables that maximize prediction capacity.

2.2 | Partial least squares allows the phenotypic prediction of the entire mutational space

Complete experimental characterization of the whole library is infeasible. We still did not determine most variants’ activity (more than 98% of the library). Hence, we applied a computational method that allows the whole library’s phenotypic prediction from the 1.8% experimentally studied. We fit the experimental data to the equation:
to-lysine in a solvent-exposed position (A39K) almost fully compensates for the functional properties of the buried Asp26.

We attempted to rationalize this result on a molecular basis. For the thioredoxin to be active, the pK$_a$ of the catalytic Cys32 must be sufficiently low to provide a convenient population of negatively charged thiolate at neutral pH.$^{40}$ In fact, removing the Asp26 increases the pK$_a$ of Cys32 (from 7.1–7.4 to 7.8–9.0), reducing its reactivity. We used PROPKA 3$^{41}$ to assess, based on the 3D structure, how the identity of the residue in position 39 influences the Cys32’s pK$_a$. To this end, we simulated the structures and performed energy minimization with Chimera 2.$^{42}$ As expected, a decrease in the Cys32’s pK$_a$ was observed when a lysine is introduced in position 39. On the contrary, an increase in the Cys32’s pK$_a$ was predicted when a glutamate is placed in position 39 (Table S2). This data must be interpreted cautiously because the predicted shift on the pK$_a$ is relatively small and differs from the experimental value. However, it qualitatively proposes an explanation for the catalytic role of the Lys39. We further use PROPKA3 to assess the Cys32’s pK$_a$ of some inactive variants (S74E/D26I-trx0 and P30E/D26I-trx0 since they are overrepresented among the inactive variants experimentally tested, and P30E/R73E/S4E/V91E/D26I-trx0 and P30E/R73M/S74E/V91E/D26I-trx0 since they are the two variants less active). As expected, for all these inactive variants, PROPKA 3 predicted an increase in the Cys32’s pK$_a$. These predictions may be limited since: (a) pK$_a$ values are challenging to reproduce based on structure-based calculations$^{43}$; (b) We lack the experimental 3D structures of the studied mutants and the structure of the catalytic cysteines in the thiolate form.

Accordingly, we simulated the structures and performed energy minimization with Chimera 2.$^{42}$

A deeper analysis of the pK$_a$ prediction allows us to further unravel how the Lys39 may shape the Cys32’s properties. According to PROPKA 3’s prediction, a coulombic interaction between Lys39 and Cys32 is one of the main contributors to the decrease in the pK$_a$ of the latter (Table S2), stabilizing the thiolate form. Further, when placing a glutamate in position 39, PROPKA 3 predicted a change in the hydrogen bond pattern between the catalytic Cys32 and Cys35 that causes an increase in the Cys32’s pK$_a$ (Table S2) that destabilizes the thiolate form. Interestingly, a lysine at position 39 is highly conserved in thioredoxins from mammalian organisms, suggesting a potential key role of this residue in modulating the pK$_a$ of Cys32 in mammals.$^{34}$

The fact that a single point mutation alanine-to-lysine in a solvent-exposed position (A39K) almost fully compensates for the functional properties of the buried Asp26 suggests a remarkable plasticity of the thioredoxin’s active site center. Further, the A39K mutation involves the accommodation of an ionizable residue in the protein surface, and thus, it likely implies a minor stability effect.

2.3 Simultaneous enhancement of redox activity and stability in thioredoxin

We showed that surface mutations could reproduce the role of a buried charge on the redox activity of thioredoxin in vitro. However, the mutations’ stability effect and the interplay between stability and activity remain to be tested.
FIGURE 3 Redesigning the conserved active site center allows the multi-feature optimization of the enzyme. (a) Most of the variants decrease the thermostability of the library's background—D26I-trx0. However, compared with trx0, half of the variants have improved stability. (b) Scanning calorimetry profiles of the variant D26I/A39K-trx0 measured in HEPES 50 mM pH 7 at three different concentrations of guanidinium hydrochloride (guanidine). Melting temperatures correspond to the maximum heat capacity of the calorimetric traces. (c) Melting temperatures of D26I/A39K-trx0 at different guanidine concentrations (open circles). Closed orange circle corresponds to the extrapolation to zero concentration of guanidine. (d) Plotting the redox activity versus the thermostability of each variant displays a spread-out distribution of the experimental dataset. (e) The PCA analysis displays two components with significant size. PC1 explained the 65% of the variance, whereas PC2 explained the 35%.
To understand the energetic landscape confined by the library variants, we studied their thermostability by differential scanning calorimetry (DSC) (Figure 3a and S2). Few variants displayed thermal stabilities equivalent to trx0. Remarkably, the A39K substitution, with similar catalytic activity to trx0, presented a $T_m$ of 137.1°C (Figure 3b, c, and d). Therefore, we wondered whether the redesigned active site center allows the multi-feature optimization of the enzyme, that is, whether simultaneously improving stability and activity is feasible.

To analyze whether stability and activity trade-off, we studied the stability-activity correlation among the experimentally studied variants. Plotting both properties shows, at a glance, a spread-out distribution of the dataset (Figure 3d). To quantify the degree of interplay between properties, we carried out a Principal Component Analysis (PCA). PCA is a mathematical algorithm that reduces the dataset’s dimensionality, preserving as much statistical information as possible. It identifies directions, called Principal Components, along which the variation in the data is maximal. If two properties correlate, the dataset can be reduced to only one component. Hence, if two properties display a major component and a very minor one. On the contrary, if two properties are independent, two components are needed to explain the data variation. Then, the PCA analysis displays two components with significant size. Two components contributed significantly to the variance of the stability-activity dataset. PC1 explained 65% of the variance, whereas PC2 explained 35% (Figure 3e). Thus, the PCA analysis revealed a weak correlation and an efficient independent modulation of both properties. Overall, the modulation range achieved for the catalysis and stability was broad and, since both parameters weakly correlate, a simultaneous improvement of activity and stability seems possible. Indeed, the mutant A39K, whose redox activity is comparable to trx0 and higher than that of E. coli and human thioredoxins, has a $T_m$ of 137°C.

Taken together, the results show that the small impact of surface mutations on protein stability permits improving the catalytic activity without compromising the stability.

### 2.4 Synthetic thioredoxin functions in vivo

The mutation A39K in thioredoxin reproduces the catalytic role of the buried Asp26 in vitro and stabilizes the protein scaffold. We wondered whether our redesigned variant is active in a cellular context. In the cell, thioredoxin must catalyze the reduction of disulfide bridges of numerous protein substrates, and multiple factors such as crowding and viscosity are at play. To assess whether the redesigned thioredoxin (D26I/A39K-trx0) functions in vivo, we genome-integrated selected thioredoxin variants (trx0, D26I-trx0, and D26I/A39K-trx0) in a Saccharomyces cerevisiae strain lacking the TRX2 gene. TRX2 encodes the protein thioredoxin II, which maintains thiol peroxidases in the reduced state. Consequently, the TRX2 deletion mutant (trx2Δ) is hypersensitive to oxidative stress. Thus, if a thioredoxin variant is active in vivo, it should, to some extent, complement the function of trx2 under oxidative stress. As a positive control, we also genome-integrated the yeast TRX2. As thioredoxin is a moonlighting protein with multiple interaction partners, a possible increase in fitness could be partially attributed to specific interactions or functionalities independent from the active site. To minimize this effect, we conduct the experiments in yeast, an organism whose thioredoxin is evolutionarily distant from trx0, which is derived from an ancestral thioredoxin corresponding to an internal bacterial node. Indeed, while trx0 and E. coli’s thioredoxin share an identity of 58.3%, the identity shared by trx0 and yeast’s trx2 is only 38.3%.

We studied the growth of these mutants in normal growing conditions and under oxidative stress. The lag time of the trx2Δ strain increased 3.3-fold and 6.3-fold in the presence of 3 mM H2O2 and 4 mM H2O2 respectively (Figures 4a, S3, and Table S3). We observed a slightly reduced lag-phase under H2O2 exposure for the D26I-trx0 variant (2.9-fold increase at 3 mM H2O2 and 5.8-fold at 4 mM H2O2), and a more reduced for trx0 (2.7-fold and 5.2-fold increase at 3 and 4 mM H2O2) and even more for the redesigned thioredoxin D26I/A39K-trx0 (2.6-fold and 5.0-fold at 3 and 4 mM H2O2; Figure 4a and Table S3). Notably, the expression of D26I/A39K-trx0 decreased the lag-phase in 3.2 hr at 3 mM H2O2 and in 9.1 hr at 4 mM H2O2 with respect to D26I-trx0. While D26I/A39K-trx0 was not able to suppress the peroxide sensitivity to the same degree than trx2 (Figure 4b and Table S3; 1.8-fold and 2.5-fold increase in the lag-phase at 3 and 4 mM H2O2 compared with the lag-phase at 0 mM H2O), it displayed similar positive complementation as trx0 and significantly improved compared with D26I-trx0. The trx0 is an ancestral thioredoxin variant that was supposed to exist in the Precambrian era. The coevolution of the thioredoxin with the organism’s protein network is fundamental to optimizing its performance. This is reflected in the superior capacity of endogenous thioredoxin (trx2) compared with D26I/A39K-trx0. However, D26I/A39K-trx0 significantly raises the redox activity in vivo with respect to trx2Δ and D26I-trx0.

The growth of the trx2Δ yeast constructs presented a lag-phase phenotype in the presence of H2O2 while no phenotype in the exponential or stationary phase.
However, stress response assays can be highly variable depending on cell growth conditions, as observed by the elongated error bars, especially at the exponential and stationary phases (Figure S3). Thus, we performed a second assay for oxidative stress induced by hydrogen peroxide treatment, optimized to obtain highly reproducible results. This plate-based spot assay allows the visualization of yeast cells' survival after treatment with 4 mM H₂O₂ for 24 hr at 30°C. trx2Δ and D26I-trx0 could not survive the peroxide exposure. Overexpression of trx0, D26I/A39K-trx0, and trx2 allows surviving the peroxide treatment. In the upper panel is shown the number of colonies before the treatment with 4 mM H₂O₂ and below after 24 hr at 30°C with YPD supplemented with 4 mM H₂O₂.

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3 | DISCUSSION

Our results illustrate that optimizing the protein’s core hydrophobicity combined with surface charge distribution’s rearrangement results in an extensive stability enhancement while maintaining the functional properties of the thioredoxin. We showed that, for a previously designed hyperstable thioredoxin, mutating a buried ionizable group results in a dramatic decrease in the redox capacity and, at the same time, increases protein stability by 10 degrees. Combining the experimental FIGURE 4 The redesigned thioredoxin, D26I/A39K-trx0, can catalyze the reduction of disulfide bonds in vivo. (a) In vivo complementation assays in yeast displayed the following scenario. The growth of trx2Δ presents a long lag-phase when stressing with H₂O₂. This lag-phase is maintained when overexpressing D26I-trx0. However, the lag-phase at the highest H₂O₂ concentration is shortened in 5 hr and 8 hr with the overexpression of trx0 and A39K/D26I-trx0, respectively. (b) Yeast strain trx2Δ complemented with yeast trx2 was grown at different H₂O₂ concentrations. The lag-phase extends as the H₂O₂ concentration grows higher, but significantly less than the rest of the complemented strains. (c) Plate-based spot assay to visualize yeast cells’ survival after treatment with 4 H₂O₂ for 24 hr at 30°C. trx2Δ and D26I-trx0 trx2Δ could not survive the peroxide stress. Overexpression of trx0, D26I/A39K-trx0, and trx2 allows surviving the peroxide treatment. In the upper panel is shown the number of colonies before the treatment with 4 mM H₂O₂ and below after 24 hr at 30°C with YPD supplemented with 4 mM H₂O₂.
screening of a library of mutations at selected positions with partial least squares reconstruction of the whole mutational space, we found out that a sole alanine-to-lysine mutation in a solvent-exposed position is enough to restore the functional properties of the buried charge. Further, we found a very weak stability and redox activity trade-off that allows thioredoxin’s multi-feature optimization. Thus, we engineered thioredoxin’s active site to efficiently catalyze the reduction of disulfide bonds in vitro and stabilize the protein scaffold 10 degrees. Lastly, we showed that the new thioredoxin could also function in vivo.

Protein engineering studies often seek improved stabilities. Several evolutionary-based strategies, such as back-to-consensus mutations and ancestral protein reconstruction, have been well used to achieve protein stabilization. Also, directed evolution techniques have shown great promise in stabilizing proteins. Overall, these methods have proven successful in stabilizing natural proteins. However, engineered proteins with the highest stabilities mostly come from de novo design. Some of them display exceptionally high thermal stabilities; for example, Top7 and DRNN have melting temperatures of 133°C and 142°C, respectively. One of the reasons for such impressive stabilities could be explained by the exclusively hydrophobic core of the designs. However, natural proteins often display functional ionizable residues in their hydrophobic cores achieving functionality at the expense of thermodynamic stability. Redesigning active sites with more stable configuration without compromising function is, thus, a challenge. Here, we have shown for the thioredoxin that redesigning its surface charge distribution nearby its active site can simulate the catalytic properties of a buried charged group. Thus, our results open a new strategy that might help design new enzymes.

In summary, this work suggests that it is possible to introduce functional innovation in a protein scaffold targeting those structural features more plastic. With this approach, we redesigned the thioredoxin active site center against conservation, expanding the evolutionary mutational space’s boundaries. We think this strategy can be generalized and exploited to stabilize natural proteins.

### 4 METHODS

#### 4.1 Sequence alignment and conservation analysis

We have carried out a BLAST (https://www.ebi.ac.uk/Tools/ssa/ncbibilast/) search using as query the E. coli thioredoxin sequence (pdb:2trx). The Swisprot database of November 2020 and the default options of the search were used. The retrieved sequences were aligned using the Smith-Waterman algorithm, and those sharing a minimum of 30% of identity and with entry names corresponding to thioredoxins (a total of 120) were retained.

The consensus logo was generated with Weblogo (https://weblogo.berkeley.edu/logo.cgi).
4.2 | Gene library construction and protein expression and purification

The combinatorial library was constructed by gene assembly mutagenesis\(^37\) and cloned into the plasmid pQE-80L with a 6xHis tag at the N-terminal to facilitate protein purification. The thioredoxin variants were expressed and purified as previously described.\(^72\) Purity of the protein fractions was assessed by SDS-PAGE.

4.3 | In vitro activity assay

In vitro redox activity of the thioredoxin variants was determined by a turbidimetric assay using insulin as a substrate.\(^35\) Briefly, we prepared protein solutions at pH 6.5 in phosphate buffer 0.1 M, 2 mM EDTA, and 0.5 mg/ml of insulin. The addition of DTT until 1 mM final concentration initiated the reaction that was monitored by measuring the absorbance at 650 nm over time. Then, the maximum value of the derivative of the kinetics was calculated. For each variant, 3 to 4 concentrations were assayed. The redox activity was quantified from the linear fit of the maximum derivative versus protein concentration (Figure 1c and S1).

4.4 | Partial least squares analysis

PLS analyses were carried out with the program Unscrambler X from CAMO software using the NIPALS algorithm (Figure 2). The dependent variable was the logarithm value of the redox activity and was auto-scaled before the analysis. We used leave-one-out cross-validation. The number of latent variables retained was the Unscrambler program’s optimum value based on the mean square error of cross-validation. The PLS analyses were carried with 10 replica sets constructed from the original set through random resampling (bootstrapping). The number of latent variables retained depends on the replica set used. Typical values were much smaller than the numbers of dependent and independent variables involved. Illustrative plots of experimental versus predicted activities are given in Figure 2c.

4.5 | Thermostability experiments

Differential scanning calorimetry experiments were performed using an automatic VP-DSC microcalorimeter (VP-Capillary DSC; MicroCal, Malvern Pananalytical) with a 0.133 ml cell volume. Solutions of 0.5 mg/ml protein concentration in 50 mM HEPES at pH 7 were warmed up from 25°C until 130°C or 140°C at 3.5 K/min scan rate. Heat up to 140°C requires manipulating the calorimeter settings under the supervision of MicroCal. For technical reasons, some experiments were performed warming up the samples to 130°C. To prevent boiling above 100°C, overpressure was implemented. All thermograms were dynamically corrected using the software provided by MicroCal. The melting temperature ($T_m$) was estimated at the maximum heat capacity of the calorimetric curve. The calorimetry curves of D26I/A39K-trx0 do not show the peak of the scan in the studied range of temperatures. Its $T_m$ was, thus, obtained heating up the protein until 130°C at a concentration of 1.6 mg/ml in 50 mM HEPES pH 7 at 2, 3, and 4 M Guanidine and extrapolating to 0 M Guanidine. The latter experiments were performed in a VP-DSC microcalorimeter (MicroCal, General Electric) at a scan rate of 1.5 K/min.

4.6 | Principal component analysis

Principal Component Analysis was performed using R\(^73\) prior to the analysis both variables (Redox activity and Thermal stability) were auto-scaled.

4.7 | Genome integration of thioredoxins in yeast

Genes encoding the ancestral thioredoxin variants trx0, D26I-trx0, D26I/A39K-trx0, and \textit{S. cerevisiae} trx2 were subcloned downstream of the yeast GPD promoter into an M3925 plasmid (trp1::KanMX3) with a cassette for genome integration.\(^74\) For genome integration, the cassette was amplified by PCR and transformed into BY4741 cells. Transformants were selected by G418 resistance, and correct integration was assessed by sequencing.

4.8 | Growth assays of yeast strains

Yeast strains were grown for 2 days in YPD containing 200 μg/ml of G418 at 30°C until reaching the stationary phase. Then, strains were diluted 1:100 in YPD with concentrations of 0, 3, and 4 mM H$_2$O$_2$ in a 96-well plate for a final volume of 150 μl per well. Three replicates of each strain were done for each condition. Strains were grown under constant shaking for 60 hr at 30°C in a plate reader. Absorbance at 600 nm was continuously monitored throughout the experiment. Analysis of the growth curves (Figure 4 and Table S3) was done with the growth rate R package.\(^75\)
4.9 Plate-based spot viability assay

Yeast cultures of 2 ml of YPD supplemented with G418 200 μg/ml were grown in a 96-well plate, at 30°C with saturated humidity until saturation. Then, from this preculture, two new cultures were prepared at 0.1 OD, the control and the oxidatively stressed culture. The control culture was prepared in YPD, and from this one, five more cultures were prepared by two times dilution. In total, for each yeast strain, six solutions in YPD with an absorbance at 600 nm of 0.1, 0.05, 0.025, 0.0125, 0.006, and 0.003, respectively. We spotted 5 μl of each culture in a YPD-agar plate, let it dry, and left it at 30°C for 2 days. Since absorbance 1 at 600 nm correspond around 0.5 × 10^7 cells/ml, we spotted 2,500, 1,250, 625, 312, 160, and 80 cells in each drop (Figure 4c). For the oxidative stress shock, we prepared six solutions in YPD supplemented with 4 mM H₂O₂ with an absorbance of 600 nm of 0.1, 0.05, 0.025, 0.0125, 0.006, and 0.003. The cultures were incubated 24 hr at 30°C with saturated humidity. Then, we spotted 5 μl of each culture in a YPD-agar plate, let it dry, and left it at 30°C for 2 days (Figure 4c).

4.10 pKₐ prediction of ionizable groups

Structure of the mutants (D26I, A39K/D26I, A39E/D26I, S74E/D26I, P30E/D26I, P30E/R73E/S74E/V91E/D26I, and P30E/R73M/S74E/V91E/D26I) was modeled on the LPBCA thioredoxin structure (pdb code: 2YJ7) with Chimera 2.42 For each mutation we chose the rotamer with the highest probability and a subsequent energy minimization was performed. Prediction of ionizable group’s pKₐ was done using the PROPKA 3.2 software41 integrated in the APBS-PDB2PQR web server (https://server.poissonboltzmann.org/).

In PROPKA 3, pKₐ values are modeled as follows:

\[ pK_a = pK_{a\text{water}} + \Delta pK_{a\text{water--protein}}. \]

where \( pK_{a\text{water}} \) is the pKₐ of the residue in water and it is a fixed value. \( \Delta pK_{a\text{water--protein}} \) is the contribution to the pKₐ from the residue’s environment and it is predicted on the basis of the supplied structure as:

\[ \Delta pK_{a\text{water--protein}} = \Delta pK_{a\text{desolv}} + \Delta pK_{a\text{HB}} + \Delta pK_{a\text{RE}} + \Delta pK_{a\text{QQ}} \]

where \( \Delta pK_{a\text{desolv}} \) is the contribution due to desolvation effects, \( \Delta pK_{a\text{HB}} \) due to hydrogen-bond interactions, \( \Delta pK_{a\text{RE}} \) due to unfavorable electrostatic reorganization energies, and \( \Delta pK_{a\text{QQ}} \) due to Coulombic interactions. \( \Delta pK_{a\text{desolv}} \) and \( \Delta pK_{a\text{RE}} \) are generic parameters and \( \Delta pK_{a\text{HB}} \) and \( \Delta pK_{a\text{QQ}} \) are calculated.

AUTHOR CONTRIBUTIONS

Maria Luisa Romero: Conceptualization (lead); formal analysis (lead); investigation (lead); methodology (lead); project administration (lead); supervision (lead); visualization (lead); writing – original draft (lead); writing – review and editing (lead). Hector Garcia Seisdedos: Formal analysis (equal); investigation (equal); methodology (equal); visualization (equal); writing – original draft (equal); writing – review and editing (equal). Beatriz Ibarra-Molero: Formal analysis (supporting); funding acquisition (lead); investigation (supporting); visualization (supporting); writing – original draft (supporting); writing – review and editing (equal).

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