Title: Combinatorial mutagenesis of N-terminal sequences reveals unexpected and expanded

stability determinants of the *Escherichia coli* N-degron pathway

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Abstract:

Although it is known that penultimate N-terminal residues can influence protein stability by

the N-degron pathway, there has not been a comprehensive effort to document nor predict these

effects for natural and synthetic sequences in prokaryotes. Here, we present the deepest

sequence coverage screen of the bacterial N-degron pathway using combinatorial site-saturation

mutagenesis, fluorescence-activated cell sorting (FACS), and next-generation sequencing (NGS).

Our results reveal new categories of exceptions and nuance to the N-degron pathway observed in

bulk and context-specific analyses. For example, we find cases where penultimate residues can

reinforce or shift the stability expected based only on the identity of position 1 (P1). We observe

lowered stability for P2-P5 motifs rich in bulky residues and Gln as well as heightened stability for

P2-P5 motifs rich in negatively charged residues, Gly, and Pro, particularly when these residues

are at P2. We find that P1 Cys can be a component of an N-degron in a sequence-specific manner.

Furthermore, we employ machine learning to predict protein stability, identifying several motifs

with unexpected fates. Our work expands the sequence determinants of the bacterial N-degron

pathway with unprecedented granularity, providing a set of sequence guidelines for proteomic

investigation and protein engineering.

Main Text:

<u>Introduction</u>

Protein degradation plays a key role in supporting natural and engineered functions of bacteria. Bacterial protein degradation pertains to human health in myriad ways, such as in the development of tools such as BacPROTACs[1] that can guide antibiotic target discovery through inducible degradation of essential proteins. Furthermore, bacterial proteins directly involved in protein degradation, such as the Clp protease system in the pathogen *Mycobacterium tuberculosis*[2,3], are attractive antibiotic targets. In applied biotechnology efforts, protein degradation tags are frequently used to attenuate hysteresis in genetic circuits that are expressed in industrial workhorse strains such as *Escherichia coli*[4–6]. Additionally, bacterial proteins that recognize and recruit N-terminal degrons (N-recognins) have also been repurposed for next-generation protein sequencing methods[7–10].

Protein degradation by the N-degron pathway, which is conserved across both prokaryotes and eukaryotes[11], can modulate protein half-life by orders of magnitude in bacteria and is therefore of significant relevance to fundamental and applied researchers. In *E. coli,* bulky (Leu, Phe, Trp, Tyr) or positively charged (Arg, Lys) amino acids at the N-terminus serve as the primary signal for N-degradation based on recognition by one or both of two N-recognins[12,13]. The adaptor protein ClpS, a prokaryotic homolog of the eukaryotic N-recognin Ubr1[14], recognizes and binds to N-degron-tagged proteins that carry bulky amino acids at the N-terminus and deposits them at the ClpAP protease, accelerating substrate turnover [15–18]. A second N-recognin, LFTR (leucyl-phenylalanyl transferase), recognizes positively charged amino acids at the N-terminus and will append one or more Leu/Phe onto the N-terminus to generate a new substrate for ClpS[19–21]. However, the identity of the N-terminal residue is not a definitive determinant of protein fate. Recent work has shown that residues downstream of the N-terminus through at least the first 5 positions of a protein can impact stability through the pathway, including situations where the fate of a protein appears to defy its N-terminal residue[18,20,22–27]. These important yet relatively limited observations strongly motivate a comprehensive analysis of N-degron sequence stability determinants in bacteria (Figure 1a). Additionally, synthetic and chemical biologists have increasingly designed alternative

protein N-termini for non-degradative purposes, such as for bio-orthogonal conjugation reactions

specific to certain N-terminal amino acids[28–31], live cell cleavage of N-terminal tags or signal

peptides, or protein ligation or excision via inteins, sortases, or subtiligases[32–36]. For these

applications and more, it is valuable to ensure that any generated neo-N-termini do not form an N-

degron.

To date, the characterization of the bacterial N-degron pathway has been performed using

low-throughput protein half-life assays, *in vitro* pathway reconstitution, and affinity purification to

screen dozens to hundreds of putative N-degrons at a time [12,19,20,24,26,37,38]. While these methods

have pioneered our understanding of N-degrons, they have their limitations. Pulldown assays are

limited to sequences found and enriched within the bacterial proteome. Low-throughput reporter

screens require extrapolation from smaller data sets, and *in vitro* experimentation can sometimes

present artifacts or other deviations from *in vivo* pathway behavior. To address these limitations,

recent advances within the protein degradation field, including exemplary work by Timms et al.[39],

have leveraged high-throughput screening and next-generation of screening N-terminal libraries

derived from the human proteome to analyze roughly 10^4 sequences within human cells. Similar

approaches have been taken to study protein degradation[40–42], but no studies to date have

screened large DNA libraries within bacteria to characterize the N-degron pathway.

To address this gap, we evaluated the role of the N-terminal 5 amino acids (P1-P5) using

degenerate oligonucleotides for combinatorial mutagenesis, reporter protein expression in wild-

type and N-recognin-deficient bacterial hosts, fluorescence-activated cell sorting (FACS), and

next-generation sequencing (NGS). From over 2 million unique sequences screened and collected,

we have mapped N-degron pathway with increased granularity, as visualized through position-by-

position substrate specificity maps and motif analysis. We additionally share a series of sequence-

based stability determinants. Specifically, at P2 Gln (destabilizing), Pro (stabilizing), and Gly

(stabilizing) show the largest single residue stability shifts. Furthermore, additive stability effects

were observed for motifs rich in bulky residues (destabilizing), negatively charged residues

(stabilizing), and flexible Gly/Ser residues (stabilizing). In an analysis of low stability sequences

without the canonically destabilizing P1 FLYWRK, we found motifs rich in P1 Cys and bulky and

negatively charged residues in P2-P5. We then trained a machine learning model, N-FIVE, on our

dataset and harnessed it to predict sequences that break canonicalstability rules. We conclude by

discussing the implications of this work on pathway specificity, N-degron design, and future

applications of our platform.

<u>Results</u>

Overview and development of an *in vivo* N-degron pathway fluorescent reporter

To monitor *in vivo* protein degradation, we adapted the ubiquitin fusion technique[43] into a

dual fluorescent reporter, creating an mCherry-ubiquitin-degron-sfGFP-His6x genetic fusion

which we cloned into a plasmid (p15a ori, CmR, P_{araBAD}). Ubiquitin (Ub), which is generally

considered inert within *E. coli*, is scarlessly cleaved at its C-terminus upon expression of Ubp1, a

Ub protease from *Saccharomyces cerevisiae*, using a second plasmid (ColE1 ori, KanR, P_{tet}). This

cleavage reveals a neo-N-terminus (neo-Nt) that is attached to sfGFP while simultaneously

producing a stable fluorescent mCherry-ubiquitin motif. The subsequent degradation of a

destabilizing motif attached to sfGFP should result in lower sfGFP:mCherry ratios for pathway

substrates in comparison to stabilizing motifs that are attached to sfGFP (Figure 1b). This dual

fluorescent reporter assay avoids various confounding effects, including variable protein

expression due to 5' ORF mRNA sequence and the amino acid sequence near the translational

start site[44–52]. We selected *E. coli* BL21 as the expression host due to its deficiency of the Lon and

OmpT proteases, helping to minimize proteolytic crosstalk and to isolate the effect of the ClpSAP

system. To further isolate the impact of each key N-recognin and control for possible changes in

reporter expression, we generated two separate variants of BL21 where either ClpS or LFTR were

inactivated through the insertion of in-frame stop codons using multiplex automatable genome

engineering[53,54]. This configuration paired with mutagenesis of neo-Nt regions serves as the basis

for our high-throughput screening of the pathway (Figure 1c).

To verify assay functionality, we screened neo-Nt sequences previously reported to be

strong and weak ClpS/N-degron pathway substrates. For positive controls, we selected reported

ClpS ligand FLFVQEL[26,55] and LVKTKASNLL, the latter derived from ClpSAP substrate Dps[20]. For

negative controls, we selected SLFVQEL, a known ClpS non-interacting sequence, and

DHGSGAWLLP[12], a reported stable motif derived from the first 10 amino acids of β-galactosidase.

Using flow cytometry, we observed distinct unimodal distributions of sfGFP/mCherry ratio across
all tested degrons and selected median sfGFP/mCherry ratio as a comparative metric. Ratiometric
comparisons revealed an average 6.0 and 5.9-fold dynamic range between SLFVQEL/FLFVQEL
and DHGSGAWLLP/LVKTKASNLL pairs, respectively (Figure 1d). We additionally screened
FLFVQEL in the ClpS⁻ host and observed a 5.2-fold ratio increase over the wild-type strain, further
evidencing that observed ratiometric changes from our assay matched expectations for N-degron
pathway mediated turnover (Figure 1e).

Deep sequence profiling reveals penultimate residue stability determinants

We next validated our N-degron screening platform on a 60-member library and profiled
the P1-P3 stability of eukaryotic-derived N-degrons containing Nt-Arg-Cys, which led us to identify
a strongly stabilizing influence of acidic residues at P3, which we verified by Western blotting[56].
After this validation on a small library, we sought to expand library size to the million-scale to
determine the stability contributions for all combinations of the first five amino acids through the
N-degron pathway. As the starting point for mutagenesis, we selected an N-degron derived from
the eukaryotic protein RAP2.2 (CGGAIISDFI)[57] due to its consistent behavior and high dynamic
range between stable and unstable arginylated variants. Using 5 consecutive NNK codons located
at the neo-N-terminus of the reporter (XXXXXISDFI, theoretical size: 3.2E6), we transformed the
library into WT and knockout strains. We then induced expression of both reporter and Ubp1 and
subsequently collected 10M gated events using fluorescence activated cell sorting (FACS). We
sorted mCherry-expressing cells into four bins that covered over three orders of GFP fluorescence
magnitude (Figure 2a). Upon analyzing 100-150M reads of an amplicon containing the
mutagenized region from the WT, LFTR⁻, and ClpS⁻ sorts, we obtained 2.29M, 2.34M, and 2.19M
unique sequences per host, respectively. To evaluate the stability of an analyzed sequence, we
then calculated the Protein Stability Index (PSI), a weighted average recently pioneered in high-
throughput protein degradation experimentation[39,58–60] . In line with expectations that most P1-P5
sequences are stable, PSI distributions were skewed towards high stability values, with over 82%
of wild-type sequences being in the top half of the PSI range (Figure 2b).

To study the sequence-specific substrate preferences of each N-recognin, we next

generated average PSI heatmaps for all amino acid and position combinations for the datasets

resulting from expression in WT (Supplemental Figure 1), LFTR⁻(Supplemental Figure 2), and ClpS⁻

hosts (Supplemental Figure 3). We observed high consistency between datasets; in a comparison

of the wild-type dataset and LFTR⁻ dataset, a linear correlation of PSI achieved an r-squared of 1.0

for the 17 amino acids that are expected to behave similarly when at P1 (Supplemental Figure 4).

As expected, Arg and Lys at P1 exhibited differential stability between LFTR⁻ and WT hosts. In

analyses of P1, Pro was excluded due to confounding effects where it is not uniformly cleaved by

Ulp1. We also examined how a protease capable of modifying some neo-Nt that contain Met at P1,

the native *E. coli* methionine aminopeptidase (MetAP), might affect assay results when the residue

at P2 is expected to be destabilizing if at P1. We compared the PSI for motifs carrying Nt-Met

(MXYYY) to those same motifs when they appear in the first four positions (XYYYZ) within the wild-

type dataset. We observed that residues that are canonically destabilizing at P1 (L,F,W,Y,R,K) had

increased mean PSI at the X-position in MXYYY motifs when compared to XYYYZ motifs

(Supplemental Figure 5). In line with the published specificity of MetAP[61,62], it did not appear that

Met-LFWYRK motifs produced low PSI N-degrons in bulk, though the aminopeptidase may be

functional on other motifs.

To study how individual amino acids impact stability as a function of position in the P1-P5

region, we computed the average PSI for each AA-position combination and plotted the differences

between WT and ClpS⁻ hosts in a heatmap (Figure 2c). In line with the reported role of ClpS as a

sequence-specific degradation initiator, the ClpS⁻ data shows sequence-agnostic parity in the

absence of the N-recognin. All cells in the corresponding P1-P5 heatmap lied within 0.17 PSI units

of each other, demonstrably smaller than the greater than 1 PSI differences observed between

canonically stable and unstable P1 residues in WT BL21 and LFTR⁻ BL21. Furthermore, P1 stability

trends match literature expectations. Notably, the PSI order for destabilizing residues is roughly

tiered as Phe = Arg < Leu < Trp = Lys = Tyr from lowest to highest average PSI. An analysis of the

100,000 lowest PSI sequences revealed an enrichment for five of these six residues (excluding

Lys) at and near P1 (Supplemental Figure 6).

Given the dominant role of P1 binding in stability determination, we next turned our focus

towards the stability contribution of P2. As P2 is positioned on the periphery of the ClpS binding

pocket in crystallized bound peptide ligands, docking studies have not clearly elucidated a role for

this position (Supplemental Figure 7). First, we observed that select residues in P2 can generate

highly destabilizing sequences in tandem with LFWYRK at P1. To this end, in an analysis of the 100

most enriched P1-P2 motifs at low PSI, 90/100 contained at least one FLWYRK residue, and 34/36

possible combinations containing two of FLWYRK residues were present, showing a clear

overrepresentation of these six residues within destabilizing motifs relative to other sequence

combinations clustered at P1-P2 (Figure 3d-3e).

We further observed that sequences containing Gln at or near P1 were amongst the most

common sequences enriched at low PSI, which has not been previously reported to the best of our

knowledge. More distinctly, Gln at position 2 (and to a lesser extent at P3-5) is amongst the most

destabilizing P2 amino acid in both WT and LFTR$^-$ BL21, providing a -0.51 PSI unit decrease at P2

in the WT dataset. This effect was amplified and significant when Gln was paired with P1 LFWYRK

($p < $1E-99, ES = 0.26), showing a 3.2x decrease in the difference in mean PSI compared to Gln

paired with an alternative P1 residue (Figure 2f). The Gln trend was visible across WT and LFTR$^-$

data and absent in the ClpS$^-$ dataset, suggesting a ClpS-mediated preference (Supplemental

Figures 1-3). Furthermore, amongst P1-P2 combinations that were enriched at low PSI but did not

carry a canonically destabilizing P1 Gln was overwhelmingly present (Supplemental Figure 8). We

sought to validate this finding using clonal examples. Upon pairing Gln with a destabilizing P1

residue we identified sequences that resulted in sfGFP:mCherry ratios that were 3.4 and 4.0 times

lower below our previous degron control (RQWYHISDFI : FLFVQEL and FQAQWISDFI : FLFVQEL,

respectively), and FQAQWISDFI showed no visible degron-sfGFP band when visualized via

Western blot (Figure 2g).

Several amino acids showed stabilizing effects at P2. We observed this most distinctly

from Gly and Pro, waning with distance from P1 with a maximal impact at position 2. Commonly

enriched in stable, PSI >3 motifs (Supplemental Figure 9), these two residues had the highest PSI

change in the (WT – ClpS$^-$) heatmap. Notably, P2 Pro paired with P1 FLWYRK led to a significant

increase 1.05-unit increase in PSI ($p < $1E-99, ES = 0.71) (Figure 2H). A similar, but smaller in

magnitude 0.63-unit increase was observed for P2 Gly (p<1E-99, ES = 0.47) (Figure 2I). Upon

testing motifs with P2 Gly, we found clonal isolates that can stabilize screened destabilizing N-

terminal amino acids in Leu and Trp, while N-terminal Arg remained destabilizing, potentially due

to Gly having an extended distance from the neo-P1 upon LFTR activity (Figure 2J).

Similar residue side chains can cumulatively amplify or dampen N-degrons in an additive manner

We next investigated the combinatorial effect of multiple amino acids, hypothesizing that

certain stability determinants could be a function of aggregate properties of amino acid sidechains,

such as hydrophobicity or charge, or otherwise influenced by local sequence context. To select

phenomena to target, we grouped amino acids based on R-group characteristics and evaluated

their enrichment in low stability sequences within the WT dataset (Figure 3a). Unstable sequences

showed an enrichment for aromatic, hydrophobic, and positively charged residues, and a

deficiency in flexible, negatively charged, and small amino acids at various P2-P5 position, but

most distinctly at P2. First, recognizing the differential enrichment in charged residues as well as

the results presented in several papers that show that a penultimate charged residue can impact

ClpSAP recognition and turnover[16,25,55], we investigated whether there could be a relationship

between the net charge of the N-terminal region and protein stability. We observed that lower PSI

sequences are more commonly found in neutral to positively charged motifs, whereas negatively

charged motifs, even with P1 FLYWRK, demonstrated large jumps in mean stability (Figure 3b). To

further investigate, we tested clonal isolates carrying P1 Phe with motifs of various charge. We

found FDEDE (-4) to be highly stabilizing and FDEAA (-2) to be lightly stabilized relative to a control

degron or to a positively charged motif in FRKAA (+2) (Figure 3c). This charge-based phenomenon

was absent in the ClpS⁻ dataset, suggesting that neutral to positive charges with P1 FLWYRK are

critical to their recognition by ClpS (Supplemental Figure 10). Additionally, in the WT dataset both

bulky (FLWY) and positively charged (RK) P1 residues showed increases in mean PSI for negatively

charged P2-P5 motifs, though P1 RK sequences demonstrated a higher increase in mean PSI

between neutral and negatively charged P2-P5 datasets (Supplemental Figure 11).

We next analyzed the impact of clustered FLWY residues in P2-P5 to inquire about the role

of multiple bulky residues in this region. For sequences carrying multiple FLWY residues at and

near the N-terminus, the average PSI was 0.56 units lower (p < 1E-99, ES = 0.45) for 2 residues

compared to a singular P1 FLWY. Comparable distributions were observed for motifs carrying 2+

FLYW residues (Figure 3d). On average, sequences carrying a single positively charged residue at

P1 exhibited a lower mean PSI than sequences with a bulky P1, though, notably, these sequences

may have multiple bulky residues appended onto P1 Arg or P1 Lys via LFTR polyaddition[19,20]

(Supplemental Figure 12). We screened several clonal isolates with multiple bulky and positively

charged residues at P1 + at least one of P2-P5 and found them all to be destabilizing (Supplemental

Figure 13).

We additionally analyzed the additive impact of multiple smaller flexible amino acids,

focusing on Gly and Ser, which frequently compose flexible N-terminal linkers. In line with the

depletion of small amino acids in destabilizing motifs, we observed an increase in PSI with

increasing Gly/Ser count downstream of a P1 FLWYRK residue (PSI increases for Gly/Ser counts

0-1: 0.09, 0-2: 0.32, 0-3: 0.64, 0-4: 0.99, all p<1E-33) (Figure 3e). As such, the presence of these

flexible motifs may aid in the functionalization and stability of N-terminal motifs with canonically

unstable residues.

N-terminal cysteine can be destabilizing under specific sequence constraints

We next investigated whether there were generalizable sequence patterns for an N-degron

with a canonically stable N-terminus (P1 not FLYWRK). To this end, we first analyzed amino acid

enrichment in sequences with moderate to low stability (PSI < 3) that did not have LFWYRK at P1.

Within these sequences, Cys and His were enriched at P1, and bulky residues (LFWY), acidic

residues (DE), and other residues (MVC) were enriched at P2-P5 (Figure 4a). In line with the

enrichment of acidic residues, we observed that canonically stable, low PSI motifs often

maintained a neutral to negative charge (Figure 4b). Interestingly, low PSI sequences with

noncanonically destabilizing P1 follow a charge preference that is opposite to that of the canonical

six destabilizing P1 amino acids, with negatively charged motifs appearing more frequently at lower

PSI. Furthermore, as previous studies have reported that bulky residues downstream of the N-

terminus can be functionally destabilizing components of an N-degron[22], we investigated

sequence patterns with multiple bulky residues paired with a canonically stable P1 residue. We

had observed that for sequences with bulky residues in P2/P3 that there was a bimodal distribution

of sequences with high and low PSI (Figure 4c). In an analysis of low PSI sequences with bulky

residues at P2 and P3, we observed that sequences with Gln, Cys, or His at P1 had the lowest

average PSIs (Supplemental Figure 14). The largest decreases in PSI were observed for canonically

stable P1 sequences with both bulky and negatively charged residues (Supplemental Figure 15).

We next turned our attention to P1 Cys to understand what downstream residues could

make this synthetically useful P1 residue an N-degron. Interestingly, P1 Cys had a lower mean PSI

(3.42) compared to other stable P1 residues (PSI 3.80), suggesting that there was a subset of

unstable sequences (Supplemental Figure 1). Upon visualizing the sequence composition of the

20,000 lowest PSI P1 Cys motifs, we observed an enrichment of bulky and negatively charged

residues, whereas positively charged and flexible residues as well as P2 Pro were depleted in these

motifs (Figure 4d). These enrichment trends aligned with residue-by-position PSI changes (Figure

4e). We generated clonal isolates of five sequences that carried candidate destabilizing motifs with

P1 Cys that had bulky and negatively charged residues in P2-P5. Two of the five candidates

presented sfGFP:mCherry ratios on par with the FLFVQEL degron, while the remaining three

offered intermediate ratio values (Supplemental Figure 16). When we performed Western blotting

analysis, at least one candidate sequence (CFYEAISDFI) was fully degraded, producing a non-

visible band on par with the strongest identified degrons (Figure 5f).

We additionally investigated other commonly functionalized P1 residues in Gly and Ser.

Both residues were notably stable at P1, though an analysis of P2-P5 residues for the 20,000

lowest PSI sequences for P1 Gly and P1 Ser revealed an enrichment in Gln, Glu, and Lys at P2-P5.

(Supplemental Figure 17).

Training and validation of N-FIVE, an N-terminal stability predictive model

To interpret the full sequence space, our next objective was to develop a stability-

predictive machine learning model. During training, we observed increased statistical performance

of our models upon applying a minimum read count threshold of 20 (RC>20). As such, we applied

this filter to trim our WT dataset to approximately 798,000 sequences prior to final model training.

To generate a base matrix, we one-hot encoded each enriched motif by amino acid and position.

We then trained an extreme gradient boosted model using XGBoost. We supplied these encodings

and the PSI from 80% of the WT dataset to generate the input for training. After hyperparameter

tuning, our model obtained an RMSE of 0.34 and an r-squared of 0.82 when validated using the

remaining 20% of data. Satisfied with the performance of our model, we dubbed our model N-FIVE

and set out to test its predictive capabilities (Figure 4a-b). We first simulated the PSI of 100,000

random N-terminal sequences and observed the characteristic decrease in mean PSI for

sequences carrying destabilizing N-termini (Figure 4c). Upon simulating the full 3.2M sequences

using N-FIVE, we observed a high correlation compared to our actual dataset (Supplementary

Figure 18). We generated SHAP plots to understand the underlying architecture, observing trends

on par with known N-degron pathway rules such as the dominant role of P1 identity

(Supplementary Figure 19).

To validate the predictive capabilities of N-FIVE, we sought to generate N-terminal motifs

with unexpected stability profiles. To this end, we sought to identify five sequences with a

canonically destabilizing P1 Leu that would have stable profiles. Excitingly, all five of the

sequences (sequence LXXXXISDFI) suggested by N-FIVE displayed stable profiles (Figure 4d). We

additionally investigated the challenging task to identify sequences that are unexpectedly unstable

with P1 residues that are not one of FLWYRK. From our five-sequence screen, one motif,

SYMMDISDFI, generated a moderately destabilizing profile (Figure 4f). SHAP values were

calculated for candidate sequences LAASYISDFI (Figure 4e) and SYMMDISDFI (Figure 4g),

revealing the cumulative effect of P2-P5 residues in overriding the P1 residue as a stability

determinant.

<u>Discussion</u>

This work presents the most comprehensive mapping of the *Escherichia coli* N-degron

pathway to date. Having collected and analyzed over 2.2M screened sequences in the P1-P5 space

within wild-type BL21, we both reinforce and provide insight into the pathway with increased

granularity. Our findings corroborate the consensus understanding of the pathway within *E. coli*,

namely that six amino acids (Leu, Phe, Tyr, Trp, Arg, Lys) when present at P1 by and large lead to

protein degradation through the ClpSAP proteolytic cascade. We have expanded this to

demonstrate how residues in the P2-P5 positions can augment or dampen P1 effects, in rare cases

even fundamentally altering the expected level of protein stability. This is most evident for the

destabilizing P2 Gln and the stabilizing P2 Pro and Gly. Notably, we observe that most destabilizing

N-terminal motifs frequently have multiple bulky and positively charged residues in P1-P5.

Exceptions to this rule frequently carry P1 Cys and His and have bulky or negatively charged

residues in P2-P5. It is unclear how P2-P5 bulky amino acids may be causing instability as there is

conflicting evidence about whether bulky residue downstream of P1 can be bound directly in the

hydrophobic pocket of ClpS[22]. It is possible that this instability could be caused by strengthened

interactions in the substrate channel or surface of the protein. Furthermore, in an analysis of N-

degrons with canonically stable N-termini, we observed an enrichment in Cys, Gln, and His at P1.

Amongst the most destabilizing motifs that we observed were those that could be generated

by LFTR. Multiple previous studies have identified that polyaddition of Leu and Phe, rather than

addition of a singular Leu or Phe, is a possibility.[19,20] The highly destabilizing nature of multiple

Leu/Phe-Arg residues in tandem coincide with previous observations that the turnover of LR-

substrates is greater than just L-substrates.[16] To this end, others have suggested that the ClpS

binding site has been optimized to accommodate LFTR-modified substrates.[63] On balance, the

generation of a highly destabilizing motif in front of Nt-Arg and Lys may routinely contribute to the

generation of some of the most highly destabilizing degrons, demonstrating the rapid turnover of

LFTR-substrates despite the requirement of an additional modification.

These findings bear relevance to the design of synthetic N-termini for applications in small

molecule or protein ligation in live cells. For example, bioconjugation efforts that rely on a P1 Cys

should avoid bulky and negatively charged residues near the N-terminus. Gly-Ser motifs are likely

to provide a stabilizing effect that would offset any degradation-promoting elements. The

overrepresentation of negatively charged residues and bulky residues in low PSI sequences with

stable N-termini and small residues in high PSI sequences is worth considering when designing

protein neo-N-termini with targeted stability levels in mind. For such designs of synthetic neo-N-

termini, the stability-predictive N-FIVE model can be consulted, including for design with sequence

constraints. Our confidence in the predictive power of N-FIVE is evidenced by its ability to solve a

complex sequence-stability problem – in this instance successfully suggesting unexpectedly

stable sequences. The identification of unexpectedly unstable sequences remains a challenge, in

part due to the small number of candidate sequences to form the dataset. Further development of

more complex ML models may improve stability prediction capabilities.

Code availability

NGS analysis code, sequence databases, and the N-FIVE model have been made available at GitHub (https://github.com/KunjapurLab/N-terminal-cluster-stability).

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Author Contributions

S.S. designed and performed most experiments, wrote the bioinformatics and machine learning code, analyzed data, and wrote the manuscript. N.C. assisted in optimizing DNA library generation. A.T. assisted in developing NGS data analysis code. K.K. assisted with cloning and assay development. A.M.K. oversaw the project, designed experiments, and assisted in writing and editing the manuscript.

Conflict of Interest Statement

A.M.K. has a financial interest in a commercial entity, Nitro Biosciences Inc. The remaining authors declare no competing interests.

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Figures:

Figure 1.

A. An overview of the *Escherichia coli* Leu/N-degron pathway. The top row shows the expected

fate for proteins with the listed amino acids at the N-terminus, where the orange and blue

shaded residues are canonically destabilizing. The left-most box illustrates the

leucyl/phenylalanyl transferase LFTR, appending an F or L onto an N-terminal R or K. The

middle box illustrates the ClpS adaptor protein that binds N-terminal F, L, W, and Y and recruits

the protein to the ClpAP protease for degradation. The right-most box shows the dual reporter

after the action of a ubiquitin (Ub) cleavase, revealing a neo-N-terminus (neo-Nt) on an sfGFP

reporter. **B.** Genetic constructs for the protein stability assay. Wild-type or knockout *E. coli*

strains were transformed to harbor two plasmids. The first plasmid harbors a synthetic gene

fusion that encodes the dual fluorescent reporter under control of an arabinose-inducible

promoter. The second plasmid harbors a gene encoding ubiquitin cleavase from

Saccharomyces cerevisiae under control of a tetracycline-inducible promoter. Constructs

were tested in wild-type BL21 as well as BL21 derivatives deficient in *clpS* and *aat* (LFTR). **C.**

The HTS protein stability workflow used in this study. It includes 10^6 sequence DNA library

generation, high-efficiency transformation, 10^7 event cell sorting, next-generation sequencing,

and data analysis that includes machine learning, quality control, stability evaluation, and motif

enrichment. **D.** The protein stability assay differentiates literature controls. sfGFP:mCherry ratio density histograms for over 100,000 events collected for cells expressing Ubp1 as well as literature control N-terminal sequences embedded in a dual fluorescent reporter construct. Stable sequences were SLFVQEL and DHGSGAWLLP and unstable sequences were FLFVQEL and LVKTKASNLL. Dual reporter ratiometric comparison of literature controls using flow cytometry. **E.** Known ClpS-binding degron FLFVQEL returns stabilizing sfGFP:mCherry ratio values in the absence of ClpS. Data shown is median fluorescence intensity ± standard deviation for at least 100,000 events captured across clonal biological triplicates using flow cytometry. Figure generated using Biorender content.

Figure 2

A. FACS binning of the XXXXXISDFI library within wild-type BL21 **B.** Violin plot showing the PSI distribution of 2.30M collected sequences from the wild-type BL21 dataset. The number of sequences rounded to the nearest thousand can be found below the corresponding PSI sub-range. **C.** Mean change in PSI heatmap between wild-type BL21 and ClpS- BL21. **D.** Log_2 enrichment ratio plot of the 400 possible stable (PSI >3)/unstable

(PSI<2) P1-P2 dipeptide motifs. Sequences containing LFWYRK in either P1 or P2 are highlighted in light blue and sequences containing LFWYRK in both P1 and P2 are highlighted in dark blue. **E.** LFWYRK composition of the 100 lowest PSI P1/P2 motifs. **F.** Mean PSIs for wild-type sequence subsets, revealing an amplified PSI lowering effect by P2 Gln when paired with a canonically unstable N-terminus. The number of sequences per condition are (++): 29,982, (+-): 754, 580, (-+): 52,021, (--): 1,459,940; (++/+-) ES: 0.26. **G.** Clonal P2 Gln isolates show decreased stability measurements relative to a literature control. Flow cytometry data for various P2 glutamine motifs (top). Western blots of hardcoded sequences (bottom) No band was visible for the FQAQWISDFI sequence. **H.** Mean PSIs for wild-type sequence subsets, revealing an amplified PSI raising effect by P2 Pro when paired with a canonically unstable N-terminus. The number of sequences per condition are (++): 33,674, (+-): 750, 888; (++/+-) ES: 0.71. **I.** Mean PSIs for wild-type sequence subsets, revealing an amplified PSI lowering effect by P2 Gly when paired with a canonically unstable N-terminus. The number of sequences per condition are (++): 41,975, (+-): 742,587; (++/+-) ES: 0.47. **J.** Clonal P2 Gly isolates show stabilizing effects for two non-LFTR modified N-termini. Flow cytometry data for various P2 glycine motifs (top). Western blots of hardcoded sequences (bottom). Comparisons are abbreviated as (P1 FLYWRK presence, P2 presence). p-values evaluated using a Mann-Whitney U test. Effect size magnitude (ES) ranges from 0 to 1, as evaluated using Cliff's Delta. * = $p < 0.05$, ES 0-0.25; ** = $p < 0.05$, ES 0.25-0.50, *** = $p < 0.05$, ES 0.50-0.75, **** = $p < 0.05$, ES 0.75-1.0

Figure 3.

A. Low stability motif heatmap depicting $\log_2$ enrichment (observed/entire dataset) of grouped amino acids. Negative = DE, Positive = RK, Hydrophobic = AVILMFWY, Polar = RNDQEKHSTYC, Aromatic = FWYH, Small = GASC, Flexible = GSDN. **B.** Boxplot showing that sequences with P1 FLWYRK and negatively charged P2-P5 show increased bulk stability. (Net charge, n, ES relative to 0 net P2-P5 charge): (-4, 35, 0.62), (-3, 1,341, 0.61), (-2, 22,647, 0.39), (-1,155,936,0.15), (0,399,998, N/A), (+1, 171, 933, 0.04), (+2, 30,185, 0.01), (+3, 2,422, 0.07), (+4, 75, 0.24). All p-values in comparison to 0 net charge data show p <0.05. **c.** Negative charge can stabilize a canonically destabilizing P1 residue. (Top) Flow cytometry data for P1-P5 sequences with varying P2-P5 charge (FDEDE: -4, FDEAA: -2, FRKAA: +2). (Bottom) Western blots of corresponding sequences. **d.** Boxplot depicting that clustered FLWY at and near the N-terminus leads to decreased stability. x-axis lists exclusive locations for FLWY in the P1-P5 motif. P1(n=206,557) compared to P1+P2 (n=54,022) shows significance (p<0.05, ES = 0.45). **e.** Boxplot showing increased PSI means and distributions for sequences with P1 FLWYRK and

448 increasing Gly/Ser residue counts in P2-P5. (Gly/Ser count, n, ES relative to 0 P2-P5 Gly/Ser):

449 (0, 477,006, N/A), (1, 257,126, 0.07), (2, 46,829, 0.25), (3, 3,505, 0.50), (4,96, 72). All p-values

450 in comparison to 0 P2-P5 Gly/Ser show p <0.05. p-values evaluated using a Mann-Whitney U

451 test, effect size magnitude (ES) ranges from 0 to 1, as evaluated using Cliff's delta. * = p <0.05,

452 ES 0-0.25; ** = p <0.05, ES 0.25-0.50, *** = p <0.05, ES 0.50-0.75, **** = p <0.05, ES 0.75-1.0

Figure 4.

A. Low stability sequences with canonically stable N-termini are rich in negatively charged and bulky amino acids in P2-P5. Weblogo depicting fold enrichment amino acids in low PSI (PSI <3) motifs that do not contain FLWYRKP at P1. Bulky residues (FLWY) depicted in blue and negatively charged residues (DE) depicted in red. **B.** PSI boxplot for sequences with varying net charge and canonically stable P1 residues shows the opposite charge trend to low stability P1 FLWYRK sequences. P1 Pro is excluded. **C.** Multiple bulky residues lower mean PSI. Violin plot of wild-type BL21 data showing PSI distributions for sequences with bulky residues at various positions. A comparison between motifs with no bulky residues (n = 1,511,962) and motifs with bulky residues only in P2 and P3 (n=56,341) shows a significant difference between mean PSI values (p < 0.05, ES = 0.25) **D.** Weblogo depicting enriched amino acids for the 20,000 lowest PSI motifs with P1 Cys. Bulky residues (FLWY) depicted in blue and negatively charged residues (DE) depicted in red. **E.** Amino acid enrichment heatmap for motifs with P1 Cys. Change in PSI is calculated for each amino acid relative to the position average. **F.** Identification of an Nt-Cys degron. (Top) flow cytometry data, (bottom) Western blots of the

471   degron-sfGFP product for candidate degrons with N-terminal cysteine. p-values evaluated

472   using a Mann-Whitney U test, effect size magnitude (ES) ranges from 0 to 1, as evaluated using

473   Cliff's delta. * = p <0.05, ES 0-0.25; ** = p <0.05, ES 0.25-0.50, *** = p <0.05, ES 0.50-0.75, ****

474   = p <0.05, ES 0.75-1.0

Figure 5.

A. Architecture of the N-FIVE machine learning model. Enriched sequences were one-hot encoded into a matrix. 80% of sequences were used to train and optimize an extreme gradient boosted machine learning model. Model validation was performed on the remaining 20% of sequences. The first 5 amino acids of control sequences (Figure 1d) were predicted using N-FIVE and found to match the expected stability profile. **B.** High stability sequences (predicted PSI >3.5) are under-enriched in P1 LFWYRK and low stability sequences (predicted PSI <2.5) are over-enriched in P1 LFYWRK. Stability exception predictions would be a challenging task.

C. PSI prediction of 100,000 randomly selected 5-amino acid sequences using N-FIVE.

Sequences with a canonically destabilizing P1 are highlighted (bulky residues in blue, positively

charged residues in orange). **D.** (Left) Predicted PSI values for five unexpectedly stable (pPSI

>3.5) candidate sequences selected with a P1 Leu constraint. (Top right) Candidate sequence

flow cytometry data, (right left) Western blots of the degron-sfGFP product. **E.** (right) SHAP

values for individual matrix contributions to the predicted PSI value of 3.73 for the candidate

sequence LDRNS. **F.** Predicted PSI values for five unexpectedly unstable (pPSI < 2.5) candidate

sequences selected with a P1 $\notin$ FLWYRK constraint. (Top right) Candidate sequence flow

cytometry data, (bottom right) Western blots of the degron-sfGFP product. **G.** (right) SHAP

values for individual matrix contributions to the predicted PSI value of 1.97 for the candidate

sequence SYMMD.

Methods

Strains, plasmid cloning, and transformations

Escherichia coli strains and plasmids used are listed in Table S1 and Table S2, respectively.

clpS and *aat* inactivations were performed as described in the Genetic knockout of N-recognins

section. Genes were purchased as G-Blocks or gene fragments from Integrated DNA Technologies

(IDT) or Twist Bioscience and were optimized for E. coli K12 using the IDT Codon Optimization Tool.

A version of scUbp1 without the N-terminal region was used for improved expression in *E. coli*[64].

Essential genetic sequences and primers can be found in Table S3 and Table S4, respectively.

Cloning amplicons were generated using KOD XTREME Hot Start polymerase and corresponding

buffering reagents. Amplicons were verified and purified by running through a 1% agarose gel for

200 V for 20 minutes followed by gel excision and extraction. Samples were assembled in

homemade Gibson assembly aliquots run at 50 °C for 30 minutes. Assemblies were transformed

into *E. coli* cloning strains DH5α or DH10β using either electroporation or chemical heat shock.

Following a one-hour outgrowth after transformation, cells were plated on permissive antibiotic

plates and were incubated at 37 °C overnight. The following afternoon, multiple colonies were

picked and incubated overnight in separate culture tubes containing LB media (10 g/L tryptone, 5

g/L sodium chloride, 5 g/L yeast extract) and the permissive antibiotic. The next day, cells were

stocked 1:1 with 30% glycerol at negative 80 °C for future usage. The remaining cells were

miniprepped for sequencing. Plasmid sequences were verified through a mixture of Sanger

sequencing through Eurofins, Genewiz, and Azenta and full plasmid sequencing through

Plasmidsaurus. Plasmids were transformed into BL21 and BL21 derivatives using either

electroporation or chemical heat shock. Two plasmid systems were transformed in series.

Materials and chemicals

The following compounds were purchased from MilliporeSigma (Burlington, MA, USA):

phosphate-buffered saline (PBS), kanamycin sulfate, glycerol, 25 nm membranes (VSWP02500)

and KOD XTREME Hot Start polymerase (Millipore 71975-3). D-glucose was purchased from TCI

America (Portland, OR, USA). Agarose, Laemmli SDS sample reducing buffer, and ethanol were

purchased from Alfa Aesar (Ward Hill, MA, USA). Anhydrotetracycline (aTc) was purchased from

Cayman Chemical (Ann Arbor, MI, USA Methanol, sodium chloride, LB Broth powder (Lennox), LB

Agar powder (Lennox), Amersham ECL Prime chemiluminescent detection reagent, and Thermo

Scientific™ Spectra™ Multicolor Broad Range Protein Ladder were purchased from Fisher Chemical

(Hampton, NH, USA). Taq DNA ligase was purchased from GoldBio (St. Louis, MO, USA). Phusion

DNA polymerase and T5 exonuclease were purchased from New England BioLabs (NEB) (Ipswich,

MA, USA). SybrSafe DNA gel stain was purchased from Invitrogen (Waltham, MA, USA). E. cloni

10G Supreme electrocompetent cells were purchased from Biosearch Technologies. Miniprep kits

were purchased from Zymo. 4-20% precast protein gels were purchased from Bio-Rad. Flow

cyometry filter caps (Chemglass Life Sciences CLS4380009), arabinose, DpnI enzyme

(FERFD1704) and Immobilon-E Western blot membranes were obtained from Fisher Scientific.

HRP-conjugated anti-6*His antibody (Proteintech HRP-66005) was obtained from Proteintech

(Rosemont, IL, USA).

Genetic knockout of N-recognins

Multiplex automatable genome engineering (MAGE) was used to inactivate the endogenous

aat (LFTR) and *clpS* genes[65]. MAGE oligonucleotides were designed using MODEST[66] to insert three

in-frame stop codons into the gene of interest. Freshly made electrocompetent cells were

resuspended in 5 μM oligonucleotide and subsequently electroporated to enable cell permeation

of the oligonucleotide. Cells were outgrown in LB media for an hour prior to plating on permissive

plates. To verify the desired genetic knockouts, allele-specific colony PCR was performed using

KAPA 2G Fast HotStart polymerase (Roche KK5801). Additional Sanger sequencing was

performed to verify asPCR hits.

Fluorescence analysis of degrons

Overnights made from biological triplicate colonies of BL21 and recognin-deficient BL21

derivatives carrying both the dual reporter and Ubp1 expressing plasmids were inoculated at a

1:100 ratio in fresh LB media containing 0.2% arabinose, 100 ng/mL aTc, 20 μg/mL

chloramphenicol, and 15 μg/mL kanamycin. Cells were grown at 37 °C for 18 hours prior to

analysis. Cells were then filtered using 35 μm filter caps into cytometry tubes containing fresh PBS

at a roughly 1:250 ratio of cell culture: PBS. Cells were analyzed on a NovoCyte Flow Cytometer

(Agilent Technologies). Prior to analysis, scatter gates were used to isolate singlet bacterial cells

of the appropriate morphology. sfGFP fluorescence was analyzed using a 488 nm laser and a

530/30 nm bandpass filter and mCherry fluorescence was analyzed using a 488 nm laser and a

660/20 nm bandpass filter. For all analyses, at least 100,000 events that passed through the

singlet gate were collected.

Western blotting of hexahistidine-tagged reporter proteins

1 mL of overnighted culture containing expressed dual reporter and Ubp1 protease were

lysed using glass beads for 15 minutes. The sample was transferred to a fresh microcentrifuge tube

and centrifuged at 12,000 RPM for 10 minutes. Subsequently, the supernatant was collected.

Sample concentration was evaluated using a Bradford assay on a SpectraMax i3x with a BSA

calibration curve. Sample concentration was normalized to 0.1 mg/mL in water with 1x SDS PAGE

loading dye and denatured at 95 °C for 10 minutes. To separate proteins by size, 10 uL of sample

is loaded into a 4–20% Mini-PROTEAN TGX gel and run at 180 V for 35 minutes. Subsequently, the

gel is transferred onto an Immobilon membrane through wet transfer overnight (25V for 15 hours

on ice). The following day, the membrane is blocked for at least one hour with 5% milk in TBST

buffer. An HRP-conjugated anti-His6x antibody is added to the solution at a 1:10,000 dilution and

is allowed to incubate at room temperature for 1 hour. Following three washes in TBST, ECL Prime

chemiluminescent reagent is added to the membrane. Following a 10 second incubation,

chemiluminescent images at varying intensities are captured using an Azure c280.

DNA library preparation

DNA library plasmids were designed to be assembled from two separate amplicons.

Primers containing 1-5 consecutive NNK codons were ordered from IDT, using a

25A/25C/25G/25T (N) or 50G/50T (K) hand-mixed nucleotide split to maximize diversity in the

mutagenized region. PCR amplicons were generated using KOD XTREME polymerase using roughly

200 ng of plasmid template per reaction with an annealing temperature of 59 °C and extension time

of 2 minutes for inserts and 3 minutes for the backbone. To maximize sequence diversity and avoid

overamplification, 5 separate PCR amplicons were generated at a low cycle count (20 cycles) and

high volume (50 μL per reaction). Amplicons were then digested using DpnI to remove the

template. The amplicons were then isolated on a 1% agarose gel and subsequently purified using
gel extraction. Amplicons were pooled and assembled using an isothermal Gibson assembly
method with a total of 300 ng of total DNA at a 1:1 molar ratio of insert: backbone. Assemblies
were then dialyzed on a 25 nm membrane placed over deionized water for 60 minutes. Five
separate assemblies were pooled and then electroporated into E. cloni 10G Supreme
electrocompetent cells. Immediately following electroporation, cells were mixed with 500 µL of
SOC media and were placed in a culture tube in a 250 RPM shaking incubator set to 37 °C.
Following a 30-minute outgrowth, cells were plated at serial dilutions to verify transformation
efficiency. For 20^5 member libraries, a transformation efficiency of 10^8 colonies/transformation
was the minimum threshold for further experimentation. Following a second 30-minute outgrowth,
5 mL of SOC media was added to the tube and cultures were allowed to grow overnight.

The following morning, library stocks were made by mixing 500 µL of undiluted culture at a
1:1 ratio with 30% glycerol, which were then stored at -80 °C. The remaining 4-4.5 mL of culture
was miniprepped using a Zymo ZR Plasmid Miniprep-Classic kit. BL21 and recognin-deficient BL21
derivatives carrying a Ubp1-expressing plasmid were made electrocompetent using standard
molecular biology procedures, concentrating 10 mL cultures to 200 µL competent cell aliquots.
Miniprepped libraries were electroporated into freshly made competent cells and followed parallel
outgrowth, efficiency measurements, overnight growth, and freezer stocking to original library
transformations. A transformation efficiency of 10^8 was obtained for all conditions. Library diversity
was confirmed prior to sorting using 2.5 Mbp next-generation sequencing through Azenta's
AmpliconEZ platform.

Cell culturing and fluorescence activated cell sorting

Overnights of BL21 and recognin-deficient BL21 derivatives carrying both library plasmids
and Ubp1-expressing plasmids were inoculated at a 1:100 ratio in fresh LB media containing 0.2%
arabinose, 100 ng/mL aTc, 20 µg/mL chloramphenicol, and 15 µg/mL kanamycin. Cells were
grown at 37 °C for 18 hours prior to analysis. Cells were then filtered using 35 µm filter caps into
cytometry tubes containing fresh PBS at a roughly 1:250 ratio of cell culture: PBS. Cells were
sorted on a BD FACS Aria Fusion using a 70 µm or 85 µm nozzle and a neutral density 1.0 filter.
Prior to analysis, scatter gates were used to isolate singlet bacterial cells of the appropriate

morphology. sfGFP fluorescence was analyzed using a 488 nm laser and a 530/30 nm bandpass

filter and mCherry fluorescence was analyzed using a 561 nm laser and a 610/20 nm bandpass

filter. Sorting conditions and sample concentrations were optimized to obtain event rates on the

order of 1,000 – 5,000 events per second and a sort efficiency >95%. 10 million gated cells were

sorted into four equally sized bins covering the sfGFP range, excluding a small gap at the bottom of

the distribution that was observed to have a higher concentration of debris and GFP-inactivating

mutants. Cells were recovered overnight in a mixture of 1:1 recovered cells: LB media containing

1% glucose. The subsequent day, cells were stocked and miniprepped using methods described

in the DNA library preparation section.

Next generation sequencing sample preparation

250 base pair (AVITI) and 405 base pair (Amplicon EZ) DNA amplicons for next-generation

sequencing were designed to contain a 5' and 3' sequencing adapter, a 3 base pair barcode for

demultiplexing, and a centralized mutagenized region to ensure coverage by both paired-end

reads. The full sequence can be found in our corresponding methods publication[56]. Amplicons

were generated using KAPA 2G polymerase, minimizing cycle counts (20 was default) to decrease

over-enrichment of select sequences. The amplicons were then isolated on a 1% agarose gel and

subsequently purified using gel extraction. Samples were then submitted to Azenta's Amplicon EZ

service (~100,000-300,000 2x250 paired end reads) or to the UD Sequencing & Genotyping Center

(~100,000,000 - 150,000,000 2x150 paired end reads on an Element AVITI).

Next generation sequencing analysis

NGS data was first converted from FASTQ.gz format to FASTA for ease of analysis. When

necessary, reads were demultiplexed using the three base pair barcodes that corresponded to

unique sorted bins. Reads were compared to the consensus sequence for the submitted amplicon.

Any reads that varied from the consensus sequence (excluding the barcode and mutagenized

region) or those that contained stop codons (TAG, TAA, TGA) in the mutagenized region were then

excluded from subsequent analysis. The mutagenized region was identified from each read based

on a 15 base pair consensus sequence located directly after the mutagenized region and translated

into an amino acid sequence. Read counts for each 5 amino acid motifs across the four bins of any

given strain. From this data, the protein stability index (PSI) was calculated using the formula

$PSI = \sum_{i=1}^{4} R_i * i$, where R_i is the fraction of reads present in bin i[58]. Each unique sequence, its

read counts per bin, and PSI were stored in a Pandas database for subsequent analysis.

For general enrichment analysis, the counts of any given residue or motif were first measured.

To calculate enrichment differences between two populations, the average occurrence of an

amino acid or motif in the appropriate population was calculated and subsequently subtracted to

obtain a difference. To obtain a ratio, groupings were normalized to the same sample size and

subsequently divided for ratiometric comparison. Log_2 analysis was performed to simplify data

visualization.

Heatmaps, boxplots, and violin plots were visualized using the Seaborn and matplotlib

Python packages[67,68] as well as GraphPad Prism. Weblogos were generated using Logomaker[69]. All

code was developed in Python. Code, sequence databases, and N-FIVE are accessible on GitHub:

https://github.com/KunjapurLab/N-terminal-cluster-stability. Full NGS data will be made

available prior to publication.

Development of N-FIVE machine learning model

During model training, we observed that applying a 20 read minimum filter minimized RMSE

and improved R^2 for the model. Utilizing the sequence-PSI database, wild-type BL21 data were

one-hot encoded into an 798,837 x 101 matrix. The first 100 columns represented a binary

encoding of each of the 20 amino acids across 5 positions, while the final column corresponded to

the PSI value. The dataset was randomly split, with 80% of the data being used for model training

and 20% being used for model validation and evaluation. An extreme gradient boosted nonlinear

regression model was then trained using the Python XGBoost[70] package for 20 rounds.

Hyperparameter optimization was performed using Optuna[71] over 30 trials, aiming to minimize

RMSE. The following parameters were set: 'n_estimators': 986, 'max_depth': 11, 'learning_rate':

0.085, 'subsample': 0.904, 'colsample_bytree': 0.810, providing an RMSE of 0.34. The final model

was then saved as a .pkl file for future analysis. To evaluate the contribution of each matrix

component, SHAP (Shapley Additive exPlanations) values were calculated[72] for the whole model

as well as for individual sequences.

Statistical analysis

To quantify differences in PSI distributions between defined groups, we applied two non-
parametric statistical methods: the Mann-Whitney U test and Cliff's Delta for effect size
calculations. The Mann-Whitney U test was selected due to the skewed nature of PSI distribution
as well as its robustness to unequal sample sizes. We used a two-sided Mann–Whitney U test to
assess whether the PSI distributions between two groups differed significantly. Cliff's Delta was
computed to quantify the magnitude and direction of the distributional shift between compared
groups. Unlike common effect size measurements such as Cohen's d, Cliff's Delta makes no
assumptions about the underlying distribution of PSI values and remains valid for skewed,
bounded, or ordinal data. Here, Cliff's Delta represents the probability that a randomly selected
value from one group is greater than a randomly selected value from the other, minus the reverse
probability to yield an effect size bounded between −1 and +1. For clonal isolate testing,
fluorescence measurements from biological triplicates were taken for at least 100,000 gated cells
using flow cytometry. In line with common practices in the field, median fluorescence ratios +/-
standard deviation were reported.

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911 **Supplementary Materials**

912 **Supplemental Figures**

914 **Supplemental Figure 1. P1-P5 WT BL21 Heatmap** A mean PSI heatmap generated from 2.29
915 million sequences for the first five positions of a protein in wild-type BL21.

LFTR- BL21 P1-P5 Average PSI

	P1	P2	P3	P4	P5
A	3.42	3.05	2.94	2.94	2.96
C	3.20	2.93	2.86	2.88	2.93
D	3.37	3.05	3.03	2.89	2.92
E	3.35	2.90	2.97	2.86	2.85
F	2.45	2.83	2.86	2.94	2.96
G	3.44	3.38	3.10	3.12	3.15
H	3.38	2.95	2.92	2.82	2.76
I	3.42	2.98	2.92	2.93	2.92
K	3.40	2.87	2.83	2.83	2.72
L	2.59	2.94	3.05	3.09	3.07
M	3.39	2.86	2.96	3.02	2.91
N	3.40	2.95	2.92	2.87	2.82
P	2.57	3.42	3.05	2.91	2.89
Q	3.33	2.58	2.79	2.76	2.71
R	3.46	3.03	3.05	3.13	3.06
S	3.39	3.11	3.08	3.06	3.07
T	3.38	2.91	2.93	2.89	2.93
V	3.40	2.98	3.07	3.06	3.10
W	2.64	2.80	2.78	2.85	2.82
Y	2.63	2.75	2.90	2.82	2.85

917 **Supplemental Figure 2. P1-P5 LFTR- BL21 Heatmap** A mean PSI heatmap generated from 2.34

918 million sequences for the first five positions of a protein in BL21 deficient in *aat* (LFTR).

ClpS- BL21 P1-P5 Average PSI

	P1	P2	P3	P4	P5
A	3.09	3.08	3.08	3.06	3.10
C	3.11	3.08	3.08	3.08	3.10
D	3.16	3.18	3.15	3.15	3.14
E	3.12	3.12	3.14	3.15	3.14
F	3.13	3.13	3.14	3.13	3.13
G	3.06	3.09	3.07	3.08	3.07
H	3.16	3.16	3.18	3.17	3.18
I	3.14	3.15	3.16	3.14	3.13
K	3.19	3.18	3.18	3.21	3.18
L	3.04	3.05	3.05	3.05	3.05
M	3.13	3.13	3.17	3.16	3.15
N	3.17	3.16	3.16	3.19	3.20
P	3.15	3.13	3.15	3.14	3.13
Q	3.12	3.18	3.18	3.18	3.17
R	3.06	3.05	3.04	3.06	3.07
S	3.05	3.05	3.04	3.05	3.05
T	3.11	3.11	3.12	3.12	3.12
V	3.06	3.07	3.06	3.05	3.05
W	3.12	3.10	3.12	3.11	3.15
Y	3.13	3.14	3.13	3.13	3.13

921 **Supplemental Figure 3. P1-P5 ClpS- BL21 Heatmap** A mean PSI heatmap generated from 2.19

922 million sequences for the first five positions of a protein in BL21 deficient in *clpS*.

Supplemental Figure 4. P1 amino acid correlation between WT BL21 and LFTR- BL21 sorts
Correlation between PSI scores for N-terminal residues between WT BL21 and LFTR- BL21 (r =
1.00) in the 10 million event sort. Arg/Lys/Pro were excluded from regression analysis. Arg and
Lys separately highlighted in orange. Ubiquitin upstream of P1 Pro is not fully cleaved,
complicating analysis.

Supplemental Figure 5. **MX- Motifs do not show similar PSI distributions for P1 LFWYRK**

motifs. (Top) Mean PSI boxplots for residues in position X from the 10M event sort. MXYYY is in

green and XYYYZ is in orange. (Bottom) Change in PSI between MXYYY and XYYYZ datasets, with

the X-position on the y-axis and the first two Y positions in the x-axis.

Supplemental Figure 6. Bulky and positively charged amino acids are enriched in low stability sequences. Weblogo depicting the difference in enrichment for each amino acid in each of the five positions in low PSI (PSI <2) samples compared to the entire dataset.

Supplemental Figure 7. Structures of ClpS bound to peptide substrates provide limited insight into the molecular basis of substrate specificity downstream of P1. A-C: Structures of ClpS bound to Fpep with nearby ClpS residues highlighted in yellow for (A) P1, (B) P2, and (C) P3. (PDB ID: 2WA8) D-F: Structures of ClpS bound to Lpep with nearby ClpS residues highlighted in yellow for (D) P1, (E) P2, and (F) P3. (PDB ID: 2W9R). No interacting residues within 4 A were obtained for P4 and P5 of the peptides.

Supplemental Figure 8. P1/P2 Gln is noticeably enriched in low stability P1-P2 motifs that do not have P1 FLWYRK. Log_2 enrichment of P1-P2 motifs in low stability (PSI <2) motifs relative to high stability (PSI >3) motifs where FLWYRK are excluded from P1. Motifs with Gln in P1 or P2 are in orange, QQ is in red.

Supplemental Figure 9. Glycine and proline are highly enriched in high stability P1 FLWYRK sequences. A P2-P5 WebLogo showing high stability P1 FLWYRK (PSI >3) sequence enrichment relative to the entire wild-type dataset for each of the five positions studied. Glycine and proline highlighted in green.

Supplemental Figure 10. ClpS- sequences show charge parity. P1-P5 net charge heatmap for

data collected in ClpS- Bl21.

P2-P5 Net Charge (P1 ∈ RK)   P2-P5 Net Charge (P1 ∈ FLWY)

Supplemental Figure 11. Separation of charged and bulky P1 residue net charge heatmaps. All

p-values in comparison to 0 P2-P5 Gly/Ser show p <0.05. p-values evaluated using a Mann-

Whitney U test, effect size magnitude (ES) ranges from 0 to 1, as evaluated using Cliff's delta. * = p

<0.05, ES 0-0.25; ** = p <0.05, ES 0.25-0.50, *** = p <0.05, ES 0.50-0.75, **** = p <0.05, ES 0.75-

1.0

977 **Supplemental Figure 12**. **Multiple positively charged residues at and near P1 lead to low PSI**

978 **sequences.** Violin plot depicting the PSI distribution in wild-type BL21 for sequences with one or

979 more RK residues in series at and adjacent to the N-terminus.

Supplemental Figure 13. Sequences with multiple FLWYRK(Q) residues show low stability.
Flow cytometry data from the protein stability assay depicting sequences carrying multiple
FLWYRK or P2 Gln residues exhibited low sfGFP:mCherry ratios relative to literature controls.

Supplemental Figure 14. P1 Cys is the most destabilizing non FLWYRK residue when P2 and P3 are bulky. Boxplot showing PSI distributions for non-P1 FLWYRKP motifs that have P2 and P3 residues in FLWY.

Supplemental Figure 15. For WT sequences with canonically stable P1, the largest PSI drops are observed for motifs with both bulky and negatively charged P2-P5 residues. Change in PSI is relative to the 0 P2-P5 FLWY, 0 P2-P5 DE dataset (mean PSI = 3.78). All p-values relative to (0,0) are p <0.5 except cells marked with # which are p >0.05. * = ES > 0.25.

Supplemental Figure 16. Full panel P1 Cys screen reveals putative N-degrons. Flow

cytometry data collected from biological triplicates for the full panel of Nt-Cys screened

sequences.

Supplemental Figure 17. Enriched residues in low stability sequences for P1 Gly and P1 Ser.

Weblogo depicting the enrichment of amino acids relative to the whole population for the lowest

20,000 PSI sequences with P1 Gly or Ser.

N-FIVE Predicted P1-P5 PSI

1006

Real - N-FIVE P1-P5 Average ΔPSI Difference

1007

1008 **Supplemental Figure 18. N-FIVE PSI predictions.** (top) Wild-type BL21 mean PSI heatmap for

1009 all 3.2 million sequences as predicted by N-FIVE. (bottom) Change in PSI for each cell between

1010 the collected WT BL21 dataset and the N-FIVE predicted dataset.

1013

Supplemental Figure 19. Individual matrix components for N-FIVE predictions. SHAP values showing the contribution of the 20 most impactful matrix features onto the N-FIVE model.

1016

1017 **Supplemental Tables**

1018 **Supplemental Table 1. Strains**

1019

Strain	Description	Key mutations
E. coli BL21 DE3	Standard *E. coli* BL21 D3	N/A
E. coli BL21 DE3 *clpS-*	BL21 with inserted in-frame stop codons for *clpS*	V46*, L47*, Q48*
E. coli BL21 DE3 *aat-*	BL21 with inserted in-frame stop codons for *aat* (LFTR)	L20*, R21*, E22*

1020

1021 **Supplemental Table 2. Plasmids**

1022 pEVOL constructs are arabinose inducible, chloramphenicol resistant and pZE constructs are

1023 aTc inducible, kanamycin resistant unless listed otherwise.

Plasmid ID	Description	Description
pSS052	pEVOL mCherry-Ub-SLFVQEL-sfGFP-His6x	Dual reporter construct
pSS054	pEVOL mCherry-Ub-FLFVQEL-sfGFP-His6x	Dual reporter construct
pSS060	pEVOL mCherry-ArgRAP2.2_nt10-sfGFP-His6x	Dual reporter construct
pSS129	pEVOL mCherry-DPSnt6-16-sfGFP-His6x	Dual reporter construct
pSS142	pEVOL mCherry-Asp-Bgalnt10-sfGFP-His6x	Dual reporter construct
pSS168	pZE Ubp1 w TetR	scUlp1 protease construct
pSS303	pEVOL mCherry-Ub-(P1-P5NNK)-RAP2.2nt10-sfGFP	Dual reporter library with mutagenized P1-P5 residues
pSS0489	pEVOL mCherry-Ub-CFYEAISDFI-sfGFP	Dual reporter construct
pSS0490	pEVOL mCherry-Ub-CYWADISDFI-sfGFP	Dual reporter construct
pSS0491	pEVOL mCherry-Ub-CEEDAISDFI-sfGFP	Dual reporter construct
pSS0492	pEVOL mCherry-Ub-CYCEMISDFI-sfGFP	Dual reporter construct
pSS0493	pEVOL mCherry-Ub-CFYDEISDFI-sfGFP	Dual reporter construct
pSS0494	pEVOL mCherry-Ub-PYFDEISDFI-sfGFP	Dual reporter construct
pSS0495	pEVOL mCherry-Ub-IFFAEISDFI-sfGFP	Dual reporter construct
pSS0496	pEVOL mCherry-Ub-IWYKHISDFI-sfGFP	Dual reporter construct
pSS0497	pEVOL mCherry-Ub-FQKQKISDFI-sfGFP	Dual reporter construct
pSS0498	pEVOL mCherry-Ub-RQWYHISDFI-sfGFP	Dual reporter construct
pSS0499	pEVOL mCherry-Ub-FDEDEISDFI-sfGFP	Dual reporter construct
pSS0500	pEVOL mCherry-Ub-FDEAAISDFI-sfGFP	Dual reporter construct
pSS0501	pEVOL mCherry-Ub-FAAAAISDFI-sfGFP	Dual reporter construct

pSS0502	pEVOL mCherry-Ub-FRKAAISDFI-sfGFP	Dual reporter construct
pSS0504	pEVOL mCherry-Ub-GLVRGISDFI-sfGFP	Dual reporter construct
pSS0505	pEVOL mCherry-Ub-GLRLRISDFI-sfGFP	Dual reporter construct
pSS0506	pEVOL mCherry-Ub-LVRLGISDFI-sfGFP	Dual reporter construct
pSS0507	pEVOL mCherry-Ub-GSVLRISDFI-sfGFP	Dual reporter construct
pSS0508	pEVOL mCherry-Ub-HWMQKISDFI-sfGFP	Dual reporter construct
pSS0509	pEVOL mCherry-Ub-NIQMKISDFI-sfGFP	Dual reporter construct
pSS0510	pEVOL mCherry-Ub-SLWKYISDFI-sfGFP	Dual reporter construct
pSS0511	pEVOL mCherry-Ub-SYMMDISDFI-sfGFP	Dual reporter construct
pSS0512	pEVOL mCherry-Ub-IKKTQISDFI-sfGFP	Dual reporter construct
pSS0513	pEVOL mCherry-Ub-LAASYISDFI-sfGFP	Dual reporter construct
pSS0514	pEVOL mCherry-Ub-LDRNSISDFI-sfGFP	Dual reporter construct
pSS0515	pEVOL mCherry-Ub-LAVPRISDFI-sfGFP	Dual reporter construct
pSS0516	pEVOL mCherry-Ub-LGPRSISDFI-sfGFP	Dual reporter construct
pSS0517	pEVOL mCherry-Ub-LGDSVISDFI-sfGFP	Dual reporter construct
pSS0518	pEVOL mCherry-Ub-QQCKWISDFI-sfGFP	Dual reporter construct
pSS0519	pEVOL mCherry-Ub-HQAQWISDFI-sfGFP	Dual reporter construct
pSS0520	pEVOL mCherry-Ub-FQAQWISDFI-sfGFP	Dual reporter construct
pSS0521	pEVOL mCherry-Ub-WGAAVISDFI-sfGFP	Dual reporter construct
pSS0522	pEVOL mCherry-Ub-RGAAVISDFI-sfGFP	Dual reporter construct
pSS0523	pEVOL mCherry-Ub-LGAAVISDFI-sfGFP	Dual reporter construct
pSS0524	pEVOL mCherry-Ub-FLCFEISDFI-sfGFP	Dual reporter construct
pSS0525	pEVOL mCherry-Ub-RWCFLISDFI-sfGFP	Dual reporter construct

1024

 Supplemental Table 3. Important gene sequences

Genetic construct	DNA Sequence
Dual fluorescent reporter (mCherry-Ub-RCKGLAALPHS-sfGFP-His6x)	ATGGTGAGCAAGGGCGAGGaggataacatggccatcatcaaggagt tcatgcgcttcaaggtgcacatggagggctccgtgaacggccacgagttcgagat cgagggcgagggcgagggccgcccctacgagggcacccagaccgccaagctg aaggtgaccaagggtggcccccctgcccttcgcctgggacatcctgtcccctcag ttcatgtacggctccaaggcctacgtgaagcaccccgccgacatccccgacta cttgaagctgtccttcccccgagggcttcaagtggggagcgcgtgatgaacttcgagg acggcggcgtggtgaccgtgacccaggactcctcactgcaggacggcgagttca tctacaaggtgaagctgcgcggcaccaacttcccctccgacggccccgtaatg cagaagaagaccatgggctgggaggcctcctccgagcggatgtaccccgagga cggcgccctgaagggcgagatcaagcagaggctgaagctgaaggacggcggc cactacgacgctgaggtcaagaccacctacaaggccaagaagcccgtgcagc tgcccggcgcctacaacgtcaacatcaagttggacatcacctcccacaacgag gactacaccatcgtggaacagtacgaacgcgctgagggccgccactccaccg gcggcATGGACGAGCTGTACAAGCAGATTTTTGTGAAGACTT TAACAGGTAAGACGATTACCCTGGAGGTGGAGTCCTCGG ACACCATCGATAATGTAAAATCAAAAATCCAAGATAAGGAA GGAATCCCTCCAGACCAGCAACGTCTGATTTTCGCAGGTA AACAACTGGAGGATGGTCGCACGCTTTCGGACTACAACAT CCAGAAAGAATCTACCCTTCATTTGGTTCTGCGTCTGCGTG GAGGACGCTGCAAAGGACTTGCAGCACTGCCCCATTCCg catccaagggcgaggagctctttactggcgtagtaccaattctcgtagagctcgat ggcgatgtaaatggccataagttttccgtacgcggcgagggcgagggcgatgcaa ctaacggcaagctcactctcaagtttatttgtactactggcaagctcccagtacc atggccaactctcgtaactactctgacctatggcgtacaatgtttttcccgctatc cagatcacatgaagcaacatgatttttttaagtccgcaatgccagagggctatgta caagagcgcactattagctttaaggatgatggcacctataagactcgcgcagag gtaaagtttgagggcgatactctcgtaaatcgcattgagctcaagggcattgatttt aaggaggatggcaatattctcggccataagctggagtataatttcaattcccataa tgtatacattaccgcagataagcaaaagaatggcattaaggcgaattttaagattc gccataatgtggaggatggctccgtacaactcgcagatcattatcaacaaaata ctccaattggcgatggcccagtactcctcccagataatcattatctctccactca atccgtgctctccaaagatccaaatgagaagcgcgatcacatggtactcctgga gtttgtaactgcagcaggcattactcatggcatggatgagctctataagctcgagc accaccaccaccactaa

scUbp1, truncated	ATGGGGAGTGGGTCTTTCATTGCTGGGCTTGTCAACGATG
	GTAATACGTGTTTTATGAACTCGGTTCTTCAGTCCCTTGCTA
	GTAGCCGTGAACTTATGGAGTTTTTGGATAATAATGTAATCC
	GTACATATGAAGAAATTGAACAGAACGAGCACAATGAGGA
	AGGTAATGGCCAAGAGAGCGCACAAGATGAGGCAACTCA
	CAAAAAAAACACTCGCAAGGGAGGTAAGGTCTATGGGAA
	GCATAAAAAGAAATTAAACCGCAAATCTTCTAGCAAGGAA
	GACGAAGAAAAGTCGCAAGAACCAGACATTACGTTTTCGG
	TGGCGTTGCGTGATCTGCTGAGCGCATTAAATGCTAAGTAT
	TATCGCGACAAACCCTACTTTAAGACTAACTCTTTATTAAAA
	GCGATGAGCAAGTCCCCGCGCAAAAATATCTTGCTTGGGT
	ACGATCAAGAAGACGCTCAGGAATTTTTTCAAAACATTCTT
	GCGGAGTTAGAATCTAATGTCAAGTCGTTAAACACAGAAA
	AGCTTGATACTACACCGGTAGCCAAGTCCGAACTTCCAGA
	CGATGCTCTGGTTGGCCAATTAAACCTTGGTGAGGTAGGC
	ACCGTGTACATTCCCACAGAACAAATTGACCCCAATTCGA
	TTTTACATGACAAATCGATTCAAAACTTTACCCCCTTTAAAC
	TGATGACCCCGTTGGATGGGATCACGGCTGAGCGCATCG
	GCTGCCTGCAATGCGGAGAGAACGGGGGAATTCGCTACA
	GTGTTTTCAGCGGATTAAGTTTGAACCTGCCGAATGAAAAT
	ATTGGAAGCACTCTTAAACTGTCCCAGTTACTGTCCGATTG
	GTCGAAACCCGAGATTATCGAGGGTGTTGAATGCAACCGT
	TGCGCTTTAACAGCTGCGCACTCACACTTGTTTGGCCAATT
	AAAGGAGTTTGAGAAGAAACCTGAAGGCTCGATTCCCGAA
	AAACTTATTAATGCCGTAAAGGACCGCGTGCACCAGATCG
	AAGAGGTCTTGGCAAAGCCGGTTATCGACGATGAAGATTA
	TAAAAAATTGCATACTGCGAATATGGTCCGCAAGTGTTCAA
	AAAGTAAACAAATTCTTATCTCTCGTCCACCACCTTTGTTGT
	CTATTCATATCAACCGCTCTGTTTTCGACCCGCGCACCTA
	CATGATTCGCAAGAACAACTCCAAGGTTTTGTTCAAGTCAC
	GCTTGAACCTGGCACCCTGGTGCTGTGATATCAACGAAAT
	CAATCTTGACGCACGCCTTCCGATGTCGAAGAAGGAAAAA
	GCAGCTCAACAAGATTCTTCTGAAGACGAGAACATTGGCG
	GAGAGTACTATACTAAATTGCATGAACGTTTTGAGCAGGAG
	TTTGAAGATTCTGAAGAAGAGAAGGAATACGATGATGCAG
	AGGGTAATTATGCATCGCATTATAACCATACCAAGGACATC
	TCCAACTACGATCCATTGAATGGAGAAGTCGACGGTGTGA
	CTTCCGATGATGAGGATGAATACATTGAAGAGACAGACGC
	GTTGGGGAATACCATCAAAAAACGTATTATTGAACACTCCG
	ACGTGGAGAACGAAACGTGAAGGATAATGAAGAACTTCA
	GGAGATCGATAACGTTAGCTTGGATGAGCCAAAAATTAAT
	GTCGAGGACCAGCTTGAAACGAGTTCTGATGAGGAAGAC
	GTTATTCCTGCTCCACCCATCAACTACGCTCGCAGCTTTA
	GTACGGTCCCAGCGACCCCTTTAACTTACTCTTTGCGCAG

| | CGTCATCGTGCACTATGGGACTCACAACTACGGACATTAT ATTGCATTTCGCAAGTATCGTGGATGTTGGTGGCGCATCTC CGATGAGACGGTCTATGTGGTAGATGAGGCCGAAGTACTG TCAACACCGGGGGTATTTATGCTTTTCTACGAGTATGATTT CGACGAGGAGACCGGAAAAATGAAAGACGACTTAGAAGC TATCCAGAGCAATAATGAGGAAGATGACGAGAAAGAACAG GAACAGAAGGGTGTCCAGGAGCCAAAAGAATCCCAGGAG CAAGGCGAAGGCGAAGAACAAGAAGAAGGGCAAGAGCA AATGAAATTTGAGCGTACGGAGGATCATCGCGACATTTCA GGGAAGGATGTGAATTAA |

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