Distinct metabolic states of a cell guide alternate fates of mutational buffering through altered proteostasis

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Summary

Changes in metabolism can alter the cellular milieu; can this also change intracellular proteostasis? Since proteostasis can modulate mutational buffering, if change in metabolism has the ability to change proteostasis, arguably, it should also alter mutational buffering. Building on this, we find that altered cellular metabolic states in *E. coli* buffer distinct mutations. Buffered-mutants had folding problems *in vivo* and were differently chaperoned in different metabolic states. Notably, this assistance was dependent upon the metabolites and not on the increase in canonical chaperone machineries. Additionally, we were able to reconstitute the folding assistance afforded by metabolites *in vitro* and propose that changes in metabolite concentrations have the potential to alter proteostasis. Collectively, we unravel that the metabolite pools are bona fide members of proteostasis and aid in mutational buffering. Given the plasticity in cellular metabolism, we posit that metabolic alterations may play an important role in the positive or negative regulation of proteostasis.

Metabolic rewiring is a common response among different organisms to their surrounding

environment[1]. Different cell types differ in their preferred mode of metabolism in order to

harness energy and generate its required set of metabolites[2–5]. Interestingly, many of the cellular

metabolites are known to change protein folding pathways by modifying protein stability and

folding kinetics *in vitro* at high concentrations. Some of the special ones like polyphosphates[6,7]

and ATP[8,9] are known to affect cellular protein aggregation during certain stresses. It is

important to know if the change in metabolite composition of cellular milieu can alter

intracellular protein folding capacity.

If metabolites and metabolism can affect proteostasis it may have two fundamental implications.

1) Metabolism-dependent change in proteostasis may aid evolution of the whole proteome when

there is a change in an organism's niche or surrounding climate that alters its internal

environment, or when an organism undergoes a large change in metabolism[4,10]. For example,

some mutations may be rendered inactive in one metabolic state (a metabolic state is defined by

the concentration of each of the metabolites the cell accumulates), while being active in a

different one. Switching of metabolic niches may expose certain phenotypes that are hidden by

metabolism-dependent genetic buffering. 2) tissue-specific metabolic differences may predispose

cell-types to aggregate or misfold particular mutant protein, even while the protein is

ubiquitously expressed. Age-dependent change in metabolism may also render certain tissues

more prone to age-dependent aggregation[11,12]. This may have implications in explaining the late-

onset tissue specificity of aggregation associated disorders[13] which has been hard to explain with

our current understanding of proteostasis components.

Given the immediate importance of the second implication, the link between cellular milieu and

proteostasis should be investigated with human diseases in mind[14]. However, given that

metabolism is intricately linked to regulation of molecular chaperones through nutrient-signaling

(TOR, gcn2, AMPK and so on) it is difficult to pinpoint any proteostasis differences found in

eukaryotic models to cellular milieu[15–18]. It will be hard to exclude the role of canonical

molecular chaperones, autophagy or translation related mechanisms that are extensively

studied[19–21]. Although these systems are extremely important in explaining cellular quality

control, the complexity of these machineries renders unmasking the role of cellular milieu in

proteostasis a daunting task. In addition, the growth conditions for cell lines (specifically non-

cancerous and untransformed cells needed in the study) are hardly defined. This place a

limitation in understanding or altering metabolism in defined directions.

In order to test this fundamental link between metabolites and proteostasis, we needed a bare

minimum framework, an *in vivo* test tube, to test this possibility. We chose *E. coli* as it is well

characterized in terms of its metabolic network, its chaperone content and the network of

chaperones that work in the cytosol[22,23]. It also has few defined mechanisms to upregulate

chaperones. We use this model and a newly devised assay to address the possibility of

metabolism affecting intracellular protein folding.

One of the ways proteostasis can be studied is by monitoring the capacity of cells to buffer non-

synonymous mutations[24,25]. While many studies have shown mixed results in terms of buffering

by molecular chaperones- the proteins that aid in folding[26–30], remarkably little is known

regarding the role of cellular chemical milieu in proteostasis and mutational buffering.

Previously it has been shown that the addition of small molecules at large concentrations in

growth media leads to mutational buffering[25,31] . Interestingly, the type of mutations buffered

depended on the nature of the small molecule as different small molecules have different

mechanisms to aid protein folding[25]. This indicated that molecular evolution can take different

routes if different molecules are present in the growth media[32]. However, we do not understand if

the physiological concentrations of metabolites present inside the cell affect protein folding and

mutational buffering.

Osmolytes are abundant in cells and cells respond to osmotic shock by rewiring metabolism[10,33]

which allows them to accumulate compensatory osmolytes[34]. Osmolytes also influence protein

stability *in vitro*[32,35–38]. We hypothesized that change in the osmotic composition of a cell may

thus change protein folding, and hence mutational buffering. To test this, we used strains with

altered levels of certain osmolytes and monitored their potential to buffer mutations in two model

proteins. Indeed, the mutational buffering capacity was altered when the metabolite pools were

altered. Interestingly, buffering capacity of the same strain in different metabolic states was

different. In all cases, mutational buffering was only evident for mutations that impair folding,

corroborating the link between proteostasis and genetic buffering. Remarkably, the metabolites

that change along with buffering capacity were able to aid protein folding *in vitro*, suggesting a

strong link between metabolite-assisted protein folding and genetic buffering. Finally, we proved

the link between metabolic state and mutational buffering by evolving strains of *E. coli* with

enhanced osmotic tolerance. These strains showed similar altered buffering capacity as seen for

metabolically compromised cells, highlighting that the protein folding environment is different in

different metabolic states. We propose that metabolic alterations can have far-reaching

consequences on mutational buffering through their influence on proteostasis.

Results

Altered metabolite uptake leads to differences in mutational buffering

To elucidate if metabolic rewiring changes capacity to buffer mutations, we used two model

proteins- Gentamicin-acetyl transferase (Gm-R, confers gentamicin resistance)[39] and Green

Fluorescence Protein (GFP - yeast enhanced variant)[40]. These proteins met a few essential

requirements. 1) Employing these model proteins, we could monitor the activity of multiple

mutants simultaneously. 2) These proteins are non-endogenous to *E. coli* and their activity is

largely independent of endogenous *E. coli* gene regulatory network except for the proteostasis

network that takes care of its biogenesis and degradation. It ensured that altered buffering of

different mutants of the proteins in different conditions is indeed an alteration in the general

mutational buffering capacity of *E. coli* (**Figure 1**). Endogenous model proteins are attractive in

terms of their potential to unearth the native dependence on the *E. coli* environment. However,

mutational buffering for these mutants are complicated by buffering pathways that are specific to

the protein chosen (**Figure 1**). This is overcome by the use of exogenous proteins. 3) The two

chosen proteins are unique protein-folds, presumably with different folding requirements. This

enabled us to exclude fold-specific artifacts. 4) GFP was amenable to *in vitro* protein folding

studies, allowing us to reconstitute the buffering activities *in vitro* thereby helping us to delineate

the molecular mechanism of buffering.

To generate a comprehensive map of mutational buffering, we developed massively-parallel

activity assays to quantitate protein activity of a large number of mutants of the test proteins.

Since Gm-R confers resistance to Gentamicin (Gm), we modified and developed a high

throughput assay to quantitate protein activity using deep-mutational scanning[24]. We used a

Glycine doublet mutant library for this protein[25] (**Figure 2A and S1A**). For the second test

protein GFP, where we could isolate clones with different levels of GFP fluorescence, we used a
random mutant library of GFP. Quantification of the intracellular folded fraction of GFP was
done using flow-cytometry. Wt mCherry was expressed along with GFP as a bicistronic
construct to control for differences in promoter activity and plasmid number (**Figure 2B**).

To obtain *E. coli* strains with altered metabolism, we chose the strain CSH4[41] and a mutant strain
on this background deleted of Proline and Glycine-betaine uptake transporters (WG350) (see
Key resource table)[42]. We chose this mutant as it is osmosensitive as compared to CSH4 (**Figure**
S1B) suggesting an altered concentration of intracellular osmolytes; this served as an
experimental model for alteration in metabolism and osmolyte composition. To validate that
deletion of uptake-transporters altered metabolism in the mutant strain we obtained the mRNAs
differentially expressed between WG350 and CSH4 using RNA-seq; differentially expressed
mRNAs were significantly enriched for genes encoding metabolic enzymes (**Figure 2C**). To
substantiate if the strains differed in terms of metabolite concentrations, we obtained an
untargeted metabolite profile for both the strains (**Figure 2D, S1C**). The strains differed in terms
of some of the metabolic features (43 metabolites with p-value <0.05). For example, the
metabolic features corresponding to Trehalose and Trehalose-6-Phosphate increased in WG350
compared to CSH4 by ~3 and ~4-fold, respectively (**Table S1**). Proline and Betaine, although the
transporters were deleted in WG350, were only marginally lower (insignificant) in WG350
compared to CSH4 **(Table S1)**, indicating the cellular biosynthetic processes are switched on in
the absence of transporters. For Glycine-betaine we found a coherent increase in the transcripts
encoding genes for its endogenous synthesis and an increase in metabolites involved in its
synthesis (**Figure S1D**). This demonstrates that cellular metabolism is rewired significantly upon
the deletion of transporters for Proline and Glycine-betaine. Taken together metabolism in CSH4
and WG350 strains were significantly different; this provided us the platform to ask if mutational
buffering differed between these strains that differed in metabolism.

To check for mutational buffering, we transformed both the strains with Gm-R Glycine-duplet
substitution (GmR-GG) library and grew them at similar selection pressure (see star methods for
details) in the presence of Gentamicin (Gm) (**Figure S1E**). To check for mutation-specific
effects it was important to normalize Minimal Inhibitory Concentrations (MICs) of the mutants
to that of Wt Gm-R, as Wt Gm-R transformed WG350 was more sensitive to Gm than CSH4

with Wt Gm-R (**Figure S1E**). Mutant pools were sequenced and analyzed to the obtain

abundance of the different mutants in the presence or absence of selection pressure (Gentamicin

in growth medium). Enrichment scores in the presence of Gm selection were calculated to

quantitate their activity as previously published[36]. Enrichment scores (a measure of activity) of a

chosen set of partially active and inactive mutants as identified with this assay correlate well

with the semi-quantitative measurements of activity as obtained by minimum-inhibitory

concentration (MIC) of Gm in presence of each of these mutants (**Figure S1F**). While most

mutants are similarly active in the two stains CSH4 and WG350, two of the partially active Gm-

R mutants (Gm-R (I26G) and Gm-R (I27G, R28G)) were less active in WG350 than in CSH4,

indicating mutation-specific differences buffering (**Figure 2E**). Rate of translation/transcription

changes protein folding environment. We confirmed that transcription/translation was not

different between these two strains using a GFP/mCherry system that is driven by the same

promoter (**Figure S1G**). The Gm-R mutations buffered differently between strains were in the

non-conserved region of the protein (**Figure 2F)**, exposed to solvent, and flexible (**Figure 2G**).

One of the mutants was predicted to decrease the stability of the protein indicating folding

problems, however, none of them were located near the active site (**Figure 2H**). This suggested

that alteration in metabolism may affect the folding of these mutants in WG350 strains.

Remarkably, a different set of small-molecule-dependent Gm-R mutants isolated in a previous

study[25] show higher activity (w.r.t Wt Gm-R) in WG350 than in CSH4 **(Figure 2I)**. All these

mutant proteins harbored mutations in conserved residues (**Figure 2F)**, that were relatively

inflexible (**Figure 2G**). While these mutations were away from the ligand-binding site, all of

them are predicted to partially destabilize the protein (**Figure 2H**). This indicated that mutations

even in the buried, conserved residues that destabilize protein structure can be buffered

differently in the two strains.

To compare mutational buffering between WG350 and CSH4 for the second test protein with a

completely different fold, we purified a population of GFP mutants (compromised fluorescence,

Pc) that show lower GFP fluorescence compared with Wt GFP in the WT *E. coli* strain

BW25113 (referred to as BW henceforth). The mutant pool (Pc) had similar fluorescence in

CSH4 and WG350 strains (**Figure S1H**), demonstrating that there is no difference between the

strains in buffering mutations on GFP and that the difference between the strains in buffering

mutations is protein-specific. Taken together, this suggests that the ability to take up metabolites

from the medium affects metabolic network and mutational buffering. This change the spectrum
of mutations buffered in a protein-specific manner.

Different metabolic states of the same cell show differences in buffering capacity

Next, we asked if altering the metabolite pool in the same strain background changes mutational
buffering. Since osmotic shock alters the metabolite pool facilitating the accumulation of
osmotically active metabolites[28], we grew the strains WG350 and CSH4 in 350mM NaCl to
change the metabolic status of the cell. We obtained transcriptomic (**Figure 3A**) and
metabolomic shifts (**Figure 3B, S2A**) associated with osmotic shock in each of the strains.
Osmotic shock altered metabolism in both the strains (**Figure 3C, S2B**) but was routed
differently in the two strains. For example, levels of Fructose 1,6-bisphosphate increased in
CSH4 upon osmotic shock but not in WG350, contrastingly, Succinate increased in WG350 but
not in CSH4 upon osmotic stress **(Table S1)**.

Interestingly, the osmotic adaptation of CSH4 and WG350 strains led to a marked similarity in
terms of their potential to buffer di-glycine mutations in Gm-R (**Figure 3D**). Gm-R (I26G) and
Gm-R (I27G, R28G), mutants less active in WG350 than in CSH4, show similar activity in these
strains in the presence of NaCl in the growth medium. The mutants of Gm-R that showed
enhanced activity in WG350 compared to CSH4, shows a noticeably smaller difference or
similar activity in both the strains during osmotic shock (**Figure 3E**). This rule out the canonical
effect of osmotic stress-induced aggregation in affecting buffering under the conditions used
here. Notably, the buffering capacity of WG350 and CSH4 towards Gm-R was similar although
the osmotic composition of the strains in the presence of the salt is different. This indicates that
compensatory mechanisms may work through different metabolic pathways to reinstate a similar
spectrum of mutational buffering.

Since metabolite composition of WG350 and CSH4 were different in the presence of salt, and it
is known that different small molecules have protein and mutation-specific effects[32], we asked if
there could be difference in mutational buffering for a different protein. Indeed, mutational
buffering of these strains during osmotic shock is different for the GFP mutant library (although
for Gm-R it was the same) (**Figure 3F**). We could identify multiple clones of GFP that showed
enhanced fluorescence in WG350 compared to CSH4 in the presence of osmotic shock. This

clearly shows that the activity of these GFP mutants are enhanced in this altered osmotic
condition. In order to validate the observed buffering, we isolated single clones from the pool
using FACS and sequenced them. The buffered pool isolated was enriched in mutation G24C
(C1 GFP) amongst other mutations E34G, N159K, K214T (C3 GFP) and D102N, I188N (C6
GFP). (**Figure S2C**). Upon retransformation, each of these mutants exhibited similar
fluorescence in CSH4 and WG350 in normal growth media while their fluorescence increased in
osmotic stress (**Figure 3G, S2D**). Notably, these exhibited not only higher fluorescence in
WG350 than in CSH4 in the presence of osmotic shock but different mutants were buffered to
different extent in WG350 under osmotic stress. Each of these clones had at least one mutation
that mapped onto highly conserved region of the protein (**Figure 3H**), were exposed and
predicted to destabilize the protein structure (**Figure 3I**). These clones confirm that differences
in metabolism can buffer mutations in conserved regions that tend to destabilize proteins, but in a
protein specific manner in different conditions.

Mutational buffering is affected through altered proteostasis

Having obtained different clones of GFP we asked if mutational buffering has contribution from
altered proteostasis. The mutations did not map close to the fluorophore of GFP[43] (**Figure S3A**)
and *in vitro* fluorescence of the purified mutants was like that of Wt GFP (**Figure 4A**)
confirming that the mutations did not affect fluorescence of the folded proteins. Refolding
studies of the purified proteins showed that the apparent rates for refolding was unchanged for
C6 while C1 refolded with a rate ~10 fold slower and C3 refolded with a rate ~5 fold slower than
Wt GFP (**Figure 4B, S3B**). However, the native state of C1 and C6 mutants were as stable as Wt
GFP towards temperature denaturation (**Figure 4C, S3C**) while C3 was marginally destabilized,
suggesting that defect in folding may depend on either the stability of the folding intermediates
or that of the native state. This demonstrated that the mutants isolated were indeed folding-
compromised mutants, most probably resulting from energetic frustrations in the folding
landscape. Thus, differences in their *in vivo* fluorescence in different strains and conditions may
reflect the differences in proteostasis capacities between strains.

To check for *in vivo* folding defects for the isolated GFP mutants, we checked for the
degradation rates of a representative set of mutants (slow folding and buffered- C1; fast folding
and buffered- C6) vis a vis Wt GFP by stalling translation (**Figure S3D**). First, we checked if the

native state of the mutants were less stable than Wt GFP *in vivo* by monitoring fluorescence of

the mature protein after blocking translation (**Figure 4D, S3E**). The turnover of the mutants did

not differ significantly from that of Wt GFP corroborating with the *in vitro* data that the native

state of these proteins is similarly stable *in vivo* at physiological temperatures. Next, to check if

folding intermediates of the mutants are less stable than Wt GFP we followed turnover of

nascent GFP polypeptides by blocking translation after briefly inducing protein expression

(**Figure 4E**). Total GFP level was monitored using immunoblotting. All the mutants degraded

faster than Wt GFP underlining that nascent chains of these mutants are folding-compromised

under *in vivo* conditions (**Figure 4E**). This clearly demonstrate that the isolated mutants (C1 and

C6) have folding intermediates that are less stable than the folding intermediates of Wt GFP

resulting in compromised folding pathways *in vivo.*

To check if the difference in proteostasis between WG350 and CSH4 with and without osmotic

shock contributed towards differential buffering in the strains, we checked the degradation rates

for the nascently synthesized chains of GFP mutants (as described above) in WG350 and CSH4

in the presence and absence of osmotic stress. The degradation rates of the mutant GFPs in

CSH4 and WG350 were similar, however, there was a sharp decrease in degradation rates of the

C1 and C6 in WG350(S) (WG350 when grown in 350mM NaCl) compared to CSH4(S) (CSH4

when grown in 350mM NaCl) (**Figure 4F (graph), S3F (gel)**). This was not a general decrease

in degradation capacity of the cell, as sGFP - a degradation prone mutants of GFP - degraded

with similar kinetics in both the stains in the presence and absence of salt-stress (**Figure 4G

(graph), S3H (gel)**). Furthermore, the general proteases were upregulated in WG350(S) rather

than being downregulated (shown later in **Figure 5A)** arguing against a possible decrease in

degradation capacity of WG350(S).

A simple kinetic simulation of protein synthesis followed by modeling competition between

folding and quality control assisted degradation (**Figure 4H, Table S2** indicated that an increase

in folding rate would be expected to show a decrease in degradation (**Figure 4I**). Thus, the

apparent decrease in degradation rate may arise solely due to differences in the folding rate *in

vivo* in the different strains and conditions while the degradation rate remains constant (as seen

for sGFP). Taken together this indicates that the mutants isolated were indeed folding mutants

and mutational buffering differed between the strains due to their differences in proteostasis
capacities.

The mutants are dependent on chemical chaperones and not on molecular chaperones

To investigate the contribution of molecular chaperones, we checked if their levels changed in
WG350(S). The set of genes involved in maintaining protein quality control was expected to be
induced (or repressed) in WG350(S) than in WG350, CSH4(S) or CSH4. However, mRNA
levels of none of the chaperones or proteases other than DnaJ and HslU/V followed the expected
over-expression pattern (**Figure 5A**). Protein levels of the canonical chaperones, GroEL and
DnaK also did not differ significantly (**Figure 5B**).

Since we isolated specific clones, it provided us with the handle to ask if these mutants were
dependent on the canonical abundant chaperones of the cytosol. Single deletions of the abundant
canonical molecular chaperones (tig, dnaK, dnaJ, secB)[23], or the proteases (hslU/V,
clpX/P/A)[44,45], did not decrease fluorescence of these clones significantly (**Figure 5C**)
suggesting that the intracellular folding of these mutants were independent of these proteostasis
machinery. It would also mean that overexpression of DnaJ or HslU/V seen in WG350(S) at
mRNA levels may not contribute to the mutational buffering of the mutants. Coherently,
overexpressing GroEL/GroES or the DnaK/DnaJ/GrpE system in CSH4 or WG350 did not
increase the fluorescence of these mutants (**Figure 5D, S4A, S4B**), suggesting that the buffering
effect in WG350(S) is independent of the concentration of these molecular chaperones. To
mimic a global increase in chaperone levels we used other environmental stressors like heat
shock and oxidative stress that are known to increase stress-response driven chaperone
levels[23,46,47]. The fluorescence of the isolated clones that show enhanced folding in the presence
of NaCl induced osmotic stress did not fluoresce better with heat or oxidative stress (**Figure 5E**).

In vitro, GroEL/ES machinery could only marginally accelerate the rate of refolding of GFP
mutants C1 and C6, whereas DnaK/DnaJ/GrpE did not accelerate the refolding rate of the
mutants (**Figure 5F, S4C**). This confirms that these mutants are not the substrates of the
abundant chaperone machinery *in vivo* and *in vitro*. To investigate if other molecular chaperones
could assist the folding of these mutant GFPs in WG350(S) more efficiently, we studied the
physical interaction of these polypeptides chains with molecular chaperones in WG350(S). To

check this, we performed *in vivo* crosslinking and measured the amount of higher-order
complexes (possibly with components of proteostasis machinery) formed by the mutants GFPs in
CSH4 and WG350 with and without osmotic shock (**Figure 5G, S4D**). The levels were not
higher in WG350(S), indicating that the chaperone association is not altered in WG350(S). Thus,
strain-specific proteostasis differences that alter mutational buffering have contributions from
components other than molecular chaperones.

To demonstrate that the mutants buffered in different metabolic states were indeed special in
terms of their independence of chaperones and not an artifact of chaperone independence of the
model proteins chosen, we checked the chaperone dependence of the partially active (Pc)
population of GFP library in molecular chaperone deletion strains (**Figure 5H, S4E**). Indeed, the
mutant pool (Pc), as well as Wt GFP, showed decreased fluorescence indicating that Wt GFP
itself is recognized by DnaJ. Fluorescence of the mutant pool increased in *ΔdnaK*. This
suggested that DnaK can recognize and hold the mutant proteins and prevent their folding.
Additionally, we were able to isolate a GroEL/ES dependent mutant of GFP from this pool
(Sadat, Tiwari, Chakraborty, Mapa, *paper under review*), proving that the GFP-mutant pool
indeed consists of mutants that are efficiently recognized by the endogenous chaperone systems.
Thus, the mutant library has mutants that are dependent on the molecular chaperones although
these were not differently active in the different metabolic states of *E. coli* we tested. Next, we
checked if the isolated mutants C1 and C6 were efficiently recognized by the most abundant
chaperone system. Chloramphenicol chase experiment with C1 and C6 in BW and *Δdnak* and
Δlon strain (**Figure 5I, S4F**) clearly shows that the mutant proteins are recognized by the DnaK
chaperone system and is effectively degraded only in the presence of DnaK and lon. Taken
together this shows that the mutant library of GFP, as well as the mutants identified to be
differently buffered in different metabolic states, are recognized and bound efficiently by the
endogenous chaperone system. We also compared the activity of the complete Gm-R GG library
as well as two mutants (from the same library) in the *ΔdnaK* strain with their activity in BW *E.*
coli strain (**Figure S4G**). The pool of mutants, as well as the two isolated mutants, showed a
decrease in GmR activity in the absence of DnaK indicating that DnaK assists in the folding of
some of the GmR mutants. Thus, the mutant pools (GFP and GmR) tested for mutational
buffering contained mutants that were dependent on the endogenous chaperones but the

difference in activity of the selected mutants in the different metabolic states was not due to the

difference in concentrations of molecular chaperones between the two strains.

Mutant-specificity and protein-specificity of folding assistance *in vivo* are a hallmark of

chemical-chaperone mediated folding. To confirm if these mutants were indeed dependent upon

chemical chaperones for folding, we obtained their refolding rates in the presence of different

metabolites (**Figure 6A**). Many of the metabolites could act as chemical chaperones to accelerate

the refolding rates. Molecules like Aspartate, Glycine, chaperoned refolding of two of the

mutants (albeit with mutation-specific effect); C6 mutant did not show enhanced refolding rate

with any of the small molecules tested. This reiterated the mutant specific effect of different

chemical chaperones[25,32] and suggested that these small molecules could indeed facilitate the

folding of these proteins and lead to mutational buffering. Of note, the space sampled in terms of

cellular small-molecules was non-exhaustive and no combinations were tried. Importantly, some

of the small molecules (like Asp) were also able to increase the thermal stability of the mutants

specifically (**Figure S5A**) indicating that these molecules can act in a mutant specific manner to

change the energetics of folding. As a control, NaCl or KCl did not have any significant effect on

C6 or C3 indicating that salt used for osmotic shock did not affect the thermodynamics of folding

of the buffered mutants in general.

To check if the small molecule milieu of the cells can aid folding, we reconstituted *in vitro*

refolding of the selected GFP mutants in the small-molecule enriched cellular extract of WG350.

Refolding of GuHCl unfolded GFP mutants was initiated by diluting out the denaturant in the

presence of extract obtained from WG350 (**Figure 6B**). Interestingly, these mutants were

refolded to only a negligible extent in the absence of extract while folding substantially in its

presence, underlining the importance of the small molecule milieu of the cell in chaperoning

protein folding. We confirmed that these lysates were free from proteins and hence molecular

chaperones that are known to assist refolding. It is important to stress that the current

technologies of extraction limit the concentration at which these small molecules can be

extracted; they are more than a thousand-fold diluted from their physiological concentrations. It

is therefore difficult to recapitulate the full potential of this mixture and reproduce the *in vivo*

differences between strains. However, it demonstrates that the metabolite pool, even on dilution,

acts as chaperone and hence can participate in cellular proteostasis. Hence, metabolic differences

have the capacity to show altered mutational buffering in different metabolic states. This is

primarily evident for mutations that destabilize proteins and make them sensitive to osmolyte

concentrations.

Taken together, metabolic differences manifest differences in mutational buffering which may

have a significant contribution from metabolite-dependent proteostasis.

Altering cellular metabolism using metabolites changes mutational buffering

Having realized that genetic and osmotic alteration in metabolism changes the spectrum of

mutational buffering, we asked if other routes of altering metabolism may show similar changes

in buffering. Addition of excess metabolites - like amino acids- is known to rewire metabolism[48].

We added excess of the amino acids, or sugars and other metabolites to rich media to alter

metabolism and check for altered buffering. Exogenous addition of many of these compounds to

growth media (individual concentrations provided in **Table S3**) also led to enhanced folding of

some of mutants GFP in CSH4 *in vivo* (**Figure 7A**). Specifically, addition of Alanine increased

the fluorescence of both the mutant proteins as seen *in vitro* (**Figure 7B**). Further demonstrating

that the mutants isolated from the screen predominantly respond to proteostasis alterations due to

differences in metabolites, and mutational buffering can be altered by altering cellular

concentration of metabolites. Interestingly, different additives had mutant-specific chaperoning

activity *in vivo* as seen *in vitro*. This unravels the complex connection between mutational

buffering and metabolic status.

Taken together with our previous observation that these mutants are primarily dependent upon

chemical chaperones for folding, we posit a prominent role of metabolism and metabolites in

regulating differences in proteostasis capacity.

Osmotic composition determines the spectrum of mutational buffering

Is the link between metabolism and mutational buffering relevant in the context of natural

evolution? To answer this, we checked if the adaptation of an organism in a different niche (that

changes metabolism) is associated with the evolution of an altered buffering capacity; we chose

to evolve WT *E. coli* (BW25113) to adapt to high salt stress (**Figure 8A**). The strain was

continuously passaged in LB in the presence of 500mM of NaCl. We kept the evolution duration

short in order to ensure that drift mutations are minimal. Multiple strains were generated by

parallel laboratory adaptive evolution, and we obtained strains that were fitter in the presence of

osmotic stress by approximately 500 generations (**Figure 8B**). We argued that adaptation for

growth in chronic hyperosmotic stress would have fixed mutations that perturb the osmotic

composition and hence rewired the metabolic status of the cell. Genome sequencing of two of

these strains demonstrates that the each of these strains have accumulated multiple different

mutations **(Figure 8C, Table S4)** including mutations in genes involved in osmolyte synthesis

(sdaA (gluconeogenesis), fucI (fructose and mannose metabolism), otsA (trehalose metabolism)).

This indicated that the strains may have fixed certain traits associated with osmotolerance by

acquiring mutations. We confirmed that these strains indeed have a difference in metabolite pool

even when grown in the absence of osmotic shock (**Figure 8D**) we could now ask if these

metabolic differences changed the mutational buffering capacity. Remarkably, the strains tested

showed buffering for the C1 and C6 mutants of GFP that were buffered in WG350 during

osmotic stress (**Figure 8E**). The activities of the mutants were two to four-fold higher in the

different evolved strains in a mutant-specific manner though the folding of Wt GFP was not

altered in these strains (**Figure 8E**). However, *E. coli* molecular chaperones DnaK and GroEL

were not upregulated in the evolved strains compared to the unevolved BW strain as checked by

immunoblotting (**Figure 8F**) indicating that canonical hubs of proteostasis were not altered in

these evolved strains. We confirmed by chase assay that degradation of the buffered mutants was

indeed slower in strains that show buffering than in the unevolved strain (**Figure 8G (graph),**

S6B(gel)). We confirmed that slower degradation of these mutants in the evolved strain was

independent of the activity of the degradation machinery as another degradation prone mutant of

GFP (sGFP) degraded as rapidly in the evolved strain as in the unevolved strain (**Figure S6C**);

impaired degradation hence is a fallout of faster folding of the buffered GFP mutants in the

evolved strains. This indicated that a genetically different strain evolved to have altered osmotic

composition is able to buffer similar mutational variation as seen in the WG350 strain during

osmotic shock. Response to osmotic shock, or a similar metabolic state once fixed in the genome

in the absence of osmotic shock, can buffer similar mutational variations. Like GFP mutants,

mutants of Gm-R also exhibited higher activity in most of the evolved strains than the unevolved

strain (**Figure 8H, 8I**). Importantly, each of the mutants had different activities in the different

osmotolerant strains. Taken together the metabolic state of a cell is directly linked to their ability
to buffer mutational variations. Adaptation to a niche with different osmolarity changes the
spectrum of mutations buffered. This, in turn, could change the route of molecular evolution of
proteins, linking metabolism to the evolution of protein sequences through alterations in
proteostasis.

Implications

In this work we have indicated the capability of *E. coli*, a prokaryotic unicellular organism, in
affecting mutational buffering through accumulation of small molecules in the cell. This, to the
best of our knowledge, is a maiden report that specifically describes the chaperoning activity of
the small molecule component of the cytosol of *E. coli*. While this fact is important, its
regulation through osmotic shock or genetic alterations makes it even more interesting in the
context of regulation of protein folding. Regulation of protein folding capacity enables condition
specific canalization or decanalization of mutants[49–52]. It is known that perturbing growth
conditions by exogenous supplementation by osmolytes can alter mutational buffering,
specifically for mutations that affect either protein stability or protein functions[24]. Similarly,
endogenous alteration in production of osmolytes (that are part of the metabolite pool) may also
change mutational buffering. This we believe would be important when organisms shift niches
and evolve new functionalities. It is tempting to speculate that conditions that have been reported
to lead to altered Hsp90 levels during the evolution of *A. mexicanus* cave dwelling variants, may
also have led to alteration in intracellular osmolyte composition[53]. While *E. coli* is a primitive
model organism to comment on adaptive strategies and genetic buffering in higher eukaryotes
with complex developmental pathways, it surely paves way for more interesting investigations to
establish the potential link between osmotic composition, and their alterations with canalization.

Given that many of the small molecules that accumulate in cells like Arginine[32], glycerol[54–56],
amino acids[32,57,58] are reported to change protein folding kinetics and thermodynamics in vitro, it
may be the complex milieu of metabolites and not osomolytes alone that can assist or alter the
folding process in a protein specific manner. Although worked on for decades, we are yet to
understand the molecular basis of protein stabilization by different osmolytes and metabolites.
Their mechanisms seem to depend on the test protein, as well as on the specific type of problems
that mutants encounter while folding[32]. While many of these molecules (like amino acids and

sugars) [59,60] can stabilize certain proteins by decreasing flexibility of non-native states, others
may prevent formation of non-native contacts or intermolecular protein aggregation by partially
destabilizing folding intermediates. Their contribution to cellular protein folding would require
further investigation given our observations.

Alteration of metabolic states, as shown in this work provides a new avenue for alteration of
proteostasis. Use of a simple model organism serves as an *in vivo* test tube to propose and test
the hypotheses related to cellular milieu. In our case we tested that composition of cellular milieu
alters with metabolism, and has the potential to alter protein folding and proteostasis network.
This link metabolism to protein folding in prokaryotes and imply that microbial evolution may
be shaped differently when metabolism is altered either due to environmental pressure of due to
genetic changes. Further work will be required to understand if similar mechanisms are prevalent
in higher organisms.

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Declarations of Interest

The authors declare no competing interests.

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TABLES

None

METHODS

EXPERIMENTAL MODEL AND SUBJECT DETAILS

Strains, Plasmids and Proteins

E. coli strain DH5α was used for cloning, WT *E. coli* (K-12 BW25113 referred to as BW) strains

was used for expression of arabinose inducible pBAD GFP mCherry and BL21 (DE3) was used

for protein expression and purification. Protein concentrations were determined

spectrophometrically at 562 nm using BCA kit (Pierce ThermoFisher Scientific). Deletion strains

were obtained from CGSC as part of Keio collection[41,42,45].

METHOD DETAILS

Construction of mutant GFP mutant library

Mutant GFP library was made in arabinose inducible pBAD vector using random mutagenesis

approach. In the 1st step YeGFP (referred to as GFP) was amplified using GFP specific primers

and Mutazyme II polymerase (Agilent Technologies; Cat no. 200550) in order to incorporate 7-

11 mutations per kb of plasmid. In the 2nd step, the product of 1st amplification was used as mega-

primer to amplify the entire plasmid with Wt GFP (pBAD GFP mCherry) as template using

Kapa Biosystems Hifi readymix (NC02955239) for 25 cycles with both annealing and extension

at 72[O]C. The Wt copy of plasmid was digested with DpnI followed by transformation of

chemically competent DH5α cells. The colonies were scraped and plasmid was prepared in pool

to yield library of GFP mutants. The said library has a total complexity of around 10,000

mutations. The reporter is constructed such that GFP and mCherry are under same arabinose

inducible pBAD promoter in an operon to give readout of GFP according to the mutation created

on it but the mCherry readout will remain similar thus serving as an internal control for

transcription, translation and inducibility.

Screening of mutant GFP library for folding mutants responsive to osmotic stress

WT *E.coli* cells (BW) were transformed with mutant GFP library maintaining 10 fold converge

for preserving mutant library complexity. Cells were induced with 0.1% arabinose at the time of

inoculation and fluorescence was observed on BD LSR II five hours post induction at 37°C after

diluting cells in 1X PBS and incubating at 37°C for 1 hour. Fluorescence of the mutant library

was studied in a pooled manner against Wt GFP. The entire library was sorted using BD Aria III

into populations of compromised mutants (low fluorescent) and active mutants (high fluorescent)

according to the GFP fluorescence. Each of these populations was purified and plasmids

prepared. CSH4 and WG350 strains were transformed with the purified populations and

subjected to osmotic stress with 350mM NaCl. The pool of mutants buffered under osmotic

stress in WG350 was sorted and single clones of GFP were picked from here and checked for

their fluorescence in presence and absence of osmotic stress. The isolated mutants having higher

fluorescence in WG350 under osmotic stress were identified by Sanger's sequencing, also cloned

in pET SUMO under BamHI and HindIII restriction sites and purified using *E. coli* BL21 (DE3)

for further characterization.

Growth curve

696   Single colony of *E.coli* cells were inoculated in LB and grown overnight at 37°C, 200 rpm.

697   Secondary inoculations were done in LB (control) and LB containing 350mM, 500mM NaCl

698   additional in honey comb plates with temperature maintained at 37°C, 200 rpm shaking.

699   Absorbance at 600nm was measured every 30 minutes using Bioscreen C (Oy growth curve Ab

700   Ltd).

Transcriptomics

3 ml of culture grown in LB overnight at 37°C from a single colony was used to re-inoculate at

0.1% culture in 10 ml of LB and grown till OD600 reaches 0.5. The culture was mixed with

equal volume of bacteria RNA protect reagent followed by RNA isolation using Qiagen RNeasy

Mini Kit and TURBO DNA-free kit (AM1907). Quality of RNA was checked using RNA 600

Nano bioanalyzer kit and MICROB Express bacterial mRNA enrichment kit (AM1905) was used

to remove rRNA. 100 ng RNA was used to prepare library using Ion Total RNA-seq kit V2

(4475936) and Ion Xpress RNA-seq barcode 1-16 kit. Final quality check and quantification was

done using DNA HS bioanalyzer (5067-4626) and qubit HS DNA kit (Q32854). Equal amount

of each sample was pooled followed by emulsion PCR (Ion PI Hi-Q OT2 200) and sequencing

(Ion PI Hi-Q sequencing 200 kit) FastQC- tool kit was used for Data QC followed by

Trimmomatic[61] to remove low quality reads. TPM was calculated using Kallisto[62,63] followed by

identification of differentially expressed genes using EB sequencing analysis pipeline

(https://bioconductor.org/packages/release/bioc/html/EBSeq.html)[64].

Metabolomics

3ml of LB was inoculated with 0.1% inoculum from overnight grown culture and grown till

OD600~0.8 at 37°C, 200 rpm. Cells equivalent to 0.1 OD were harvested at 14000 rpm, 1 min,

4°C. Supernatant was discarded and 200ul 80% chilled methanol (80% MetOH, PIPES 1ng/µl,

U13C-U15N-glutamine in MS grade water) was added to the pellet and incubated in ice for 5

minutes for quenching. This is followed by sonication in water bath for 15 min at 4°C with

intermittent vortexing. Metabolites were collected in the supernatant by centrifugation at 14000

rpm for 1 min at 4°C. Previous step was repeated twice by adding 100µl 80% chilled methanol

each time to increase metabolite yield and stored in -80 refrigerator till further analysis. The

untargeted mass profile (or metabolic profile) was acquired in Thermo Q-exactive Orbitrap

coupled with Thermo Accucore RP C18 150*2.1, 2.6µM column with flow rate of 200µl/min.

Ion masses from 80-1000m/z was collected in negative ion mode. The raw output files obtained

from mass spec were converted into .mzXML files using proteowizard tool

(http://proteowizard.sourceforge.net/) followed by identification and analysis using

metabolomics data processing software platform XCMS and MAVEN(http://genomics-

pubs.princeton.edu/mzroll/index.php?show=index)[65,66].

Minimum Inhibitory Concentration (MIC) study of Gm-R glycine doublet mutants

The library of Gm-R glycine doublet mutants as reported previously (Bandyopadhyay *et al.*,

2012) were grown overnight in LB containing ampicillin (100µg/ml) and Arabinose (0.1%) at

37°C, 200 rpm. Secondary inoculations were done in LB containing Ampicillin (100µg/ml),

Arabinose (0.1%) and increasing concentrations of gentamicin (0-800 µg/ml) incubated for 16hrs

at 37°C and 200rpm. Growth was assessed by measuring absorbance at 600nm in flat bottom 96

well microtiter plate using TECAN infinite 200 pro. Absorbance value for each sample under

selection pressure by gentamicin was normalized against absorbance of respective unselected

sample. To obtain a semi-quantitative indication of the activity of different mutants, Wt Gm-R

transformed cells were grown as control, and the MIC for the mutants was normalized with

respect to the MIC of Wt Gm-R in the same strain.

Amplicon sequencing

Plasmids were isolated from unselected (no gentamicin) and selected cells under gentamicin

selection pressure. Gm-R gene was amplified for 25 cycles using Kapa Hifi Hotstart polymerase.

The 500bp amplicons were gel purified and quantified using qubit. 150ng of purified products

was used for library preparation Ion Plus Fragment Library Kit (part no.4471252) and Ion Xpress

Barcode Adapters (4471250, 4474009). The final library was quantified and equal amount of

DNA library of 8 samples were pooled and emulsion PCR and sequenced using Ion PGM Hi-Q

OT2 Kit (part no A29900) and Ion PGM Hi-Q sequencing kit (part no. A30044) on Ion Torrent

platform. Analysis was done using FastqQC followed by trimmomatic (cutoff Q15) and quality

check by fastQC. Good reads were aligned to Gm-R gene followed by variant calling and fitness

score calculation www.github.com/kc-lab/dms2dfe [24].

Quantitation of activity of Gm-R Glycine-doublet substitution mutants

To quantitate activity of the mutants, we obtained the read count for each of the mutant and

normalized it with respect to the coverage at the respective positions. The relative read-counts of

the mutants did not differ between the different strains in the absence of any selection pressure

for Gm-R indicating that the library was homogeneously covered in all the transformations. We

thereby did not obtain relative enrichment but compared the coverage-normalized read-counts

between the different strains and conditions to obtain differences in activity. All the amplicon-

based sequencing experiments were done in duplicates.

To obtain relative activity of the different mutants in one strain (for comparison of activity of

mutants as measured by amplicon sequencing and MIC assay as shown in **Figure S1F**) we

obtained the enrichment score for each mutant by dividing the normalized read count for each

mutant in presence and absence of Gm selection. z-scores were obtained for each of the mutant

assuming a normal distribution. A positive z-score indicates higher than average activity while a

negative z-score indicated a lower than average activity. Mutants were picked that had a high z-

score (1.4, more active than average) or low z-score (-1.6 less active than average) or one close

to zero (z-score = -0.4, activity close to average) for checking the MICs. Note: z-score was not

used to obtain the significance of difference but the mutants that exhibited either high or low

activity.

Chase for degradation of protein

From overnight grown culture in LB containing Ampicillin (100μg/ml) secondary cultures were

inoculated with 0.1% inoculum in 100 ml LB containing Ampicillin (100μg/ml) and grown till

OD_{600} 0.5 or for 5hrs (with inducer for chase of folded protein as shown in **Figure S3E**). Cells

were harvested and resuspended in 10ml of the spent media (or media containing 350mM NaCl

added in excess to LB. GFP was induced using 0.5% arabinose for 5 minutes at $37^{O}C$ and 200

rpm. Chloramphenicol (50μg/ml) was used to arrest translation. 1 ml of culture was taken out for

uninduced, 0 min, 15 min, 30 min, 60 min, 120 min post translation arrest, harvested, snap freeze

and protein was collected by lysing cells resuspended in 1X PBS using 2X lysis buffer (Tris pH

6.8 (80mM), SDS 1%, glycerol 10%). 30μg of protein estimated using Pierce™ BCA Protein

Assay Kit and loaded on SDS-PAGE followed by Western blotting probing for GFP using

Rabbit anti-GFP (Ab290) and HRP conjugated goat anti-rabbit IgG (SantaCruz Biotechnology).

In-vivo **crosslinking:**

100ml of LB medium (with and without 350mM of NaCl) containing Ampicillin (100μg/ml) was

inoculated with overnight grown cells at 0.1% inoculum for each strain. Cells were grown till

they reach $O.D_{600}$ 0.5 and harvested at 4000 rpm for 10 mins at RT. Pellet was finally

resuspended in 2ml of spent media and induced for 5 minutes at $37^{O}C$ with 0.5% arabinose for

GFP expression. Cells were harvested and resuspended in 2 ml of 1X PBS containing protease

inhibitor cocktail (Roche) and 500μl of resuspended cells was taken as a control for

uncrosslinked sample (referred to as -DSG in the text). To the remaining cells, 300μM of Di (N-

succinimidyl) glutarate (DSG) was added and incubated at $37^{O}C$ for 10 minutes for crosslinking.

Reaction was quenched with 100mM of Tris, pH 8 for 5-10 minutes at room temperature. Cells

were harvested at 13000 rpm, 2 minutes at room temperature and lysed by Freeze-thaw method

in the presence of protease inhibitors. 30μg of protein was loaded on SDS-PAGE followed by

Western Blotting probing for GFP.

Western Blotting Experiment

E. coli cells were inoculated in LB and grown at $37^{O}C$ for 5 hours. Cells were harvested and

resuspended in 1X PBS (Himedia; ML023). Equal volume of 2X lysis buffer was added and

boiled at 95°C for 15 minutes followed by spin at highest rpm for 15 minutes. The supernatant

was collected and estimated for the amount of protein. 30μg of protein was used for SDS-PAGE

and Western Blotting post transferring onto nitrocellulose membrane (Millipore; HATF00010).

The blots were decorated with antibodies against GroEL (Enzo; 9A1/2) and DnaK (Enzo; 9E2/2)

isolated from mouse. Blots were developed using HRP conjugated Goat anti-mouse IgG

(Genscript) and luminata crescendo (Millipore). Densitometric analysis was done using ImageJ.

(Rasband, W.S., ImageJ, U. S. National Institutes of Health, Bethesda, Maryland, USA,

http://imagej.nih.gov/ij/, 1997-2011).

GFP refolding assay

60μM native GFP (or mutant) was mixed with 8M GuHCl prepared in GFP refolding buffer

(Buffer-A) (Tris-Cl (25 mM), KCl (150mM), MgCl$_2$ (10mM), DTT (2 mM), pH 7.4) in 1:3 ratio.

After incubation at 25°C for 1 hour, refolding of GFP was initiated in Buffer-A by diluting the

mixture 100-fold (final GFP concentration 150 nM). Real time fluorescence of GFP was

monitored in Fluorolog 3 spectrophotometer (Horiba Jobin Yvon, with operating software

FluorEssence v3.0) at 25°C, with excitation wavelength 488 nm (2 nm slit width) and emission

wavelength 515 nm (5nm slit width), enabling 'anti-photobleaching' mode. The data acquired

from refolding assays was analyzed in OriginPro 8 (OriginLabcorporation).

Thermal melt

Buffer-A with or without 100 mM small molecules was added with purified GFP (Wt or mutant)

to a final concentration of 150nM. 20μl of this whole mixture was aliquoted in a 386 well RT-

PCR plate. The thermal melt was carried out in CFX384™ Real Time System attached to a

C1000 Touch Thermal Cycler (Biorad). The program was set to go from 25°C to 90°C with a

step of 1°C and incubation time of 2 min/°C. The data was collected as GFP fluorescence at each

step, and fit to a linear equation to obtain midpoint of each curve as melting temperature (Tm).

Isolation of protein free cell extract

E. coli K12 (BW25113) strain was grown in 5 ml LB broth at 37°C, 200 RPM orbital shaking for

overnight and this primary culture was used to inoculate a 500 ml secondary culture in LB broth.

After growing the secondary culture in identical conditions for 3 hours (OD 0.7-0.8), the culture

was centrifuged at 4000 RPM for 30 minutes at 37°C. 5 ml of boiling hot ultrapure water was

added directly to the pellet, resuspending it vigorously, and the cell suspension was immediately

collected in a glass test tube. The glass tube was kept in a boiling water bath for 30 minutes,

ensuring maximum lysis of cells and precipitation of proteins. The resulting suspension was

cooled down to room temperature and was centrifuged at 13000 RPM for 20 minutes. Absence

of protein in the supernatant, the protein free cell extract, was confirmed by Bradford protein

estimation assay.

Simulation

Numerical simulations were performed using the ODEs defined in **Figure 4** to model a basic

framework to check the apparent rate of degradation as a function of folding rate in vivo. The

simulation was set up with 1000μM S (pool of DNA/RNA that is competent to make proteins). S

can form U (nascent polypeptides) with an overall unimolecular rate constant of ktrans (a

simplistic combination of rates of transcription and translation). The pool of U can either convert

to F (folded GFP) or be degraded with the unimolecular rate constants of k_f and k_{deg},

respectively. The pool of S can be blocked by I (inhibitor of translation) with a rate constant of

k_I. This was included to mimic translation arrest by chloramphenicol. The simulation was run in

the absence of I for 300 seconds. Following this, 1mM of I was dosed into the simulation to

rapidly quench S. Following this we monitored the total concentration of uncleaved GFP (U+F)

over time to mimic the results obtained from the chase experiments performed with anti-GFP

antibodies.

Conservation score and fractional ASA measurement through FoldX

PDB structures were used to calculate the conservation score of each of the residues using

multiple sequence alignment through CONSURF[67]. Fractional ASA, and temperature factor of a

residue was obtained using VADAR[68]. FoldX package was used to obtain the predicted stability

value for each of the mutants[69]. The calculations were repeated at least 3 independent times to

obtain an average and standard deviation value for the predicated stability values.

QUANTIFICATION AND STATISTICAL ANALYSIS

Student's t test and R package for non-linear regression was used for statistical analysis. Flow-

cytometry data was analyzed using octave

DATA AND SOFTWARE AVAILABILITY

All data are provided in the manuscript. We did not develop any new software.

 KEY RESOURCES TABLE

REAGENT	SOURCE	IDENTIFIE
Antibodies		
Anti GroEL (*E.coli*) mouse monoclonal	ENZO	9A1/2
Anti DnaK (*E.coli*) mouse monoclonal	ENZO	8E2/2
Anti DnaJ (*E.coli*) rabbit polyclonal	ENZO	SPA410
Anti Tig (*E.coli*) rabbit polyclonal	Genescript	A01329
Anti GFP rabbit polyclonal	Abcam	Ab290
Anti RecA rabbit polyclonal	Abcam	Ab125096
Anti Mouse secondary	Santa Cruz Biotechnology	SC2005
Anti Rabbit secondary	Santa Cruz Biotechnology	SC2030
Di(N-succinimidyl) glutarate (DSG)	Sigma	80424
Bacterial and Virus Strains		
E.coli BL21 (DE3)		
E.coli DH5α		
BW (*E.coli* K12, BW25113) F-, DE(araD-araB)567, lacZ4787(del)::rrnB-3, LAM-, rph-1, DE(rhaD-rhaB)568, hsdR514	(Baba *et al.*, 2006)	CGSC#: 7636
Δ*Tig* (JW0426-1) *F-, Δ(araD-araB)567, ΔlacZ4787(::rrnB-3), Δtig-722::kan, λ-, rph-1, Δ(rhaD-rhaB)568, hsdR514*	(Baba *et al.*, 2006)	CGSC#: 8589
Δ*dnaK* (JW0013-4) F-, Δ(araD-araB)567, ΔlacZ4787(::rrnB-3), λ-, rph-1, ΔdnaK734::kan, Δ(rhaD-rhaB)568, hsdR514	(Baba *et al.*, 2006)	CGSC#: 8342
Δ*dnaJ* (JW0014-1) F-, Δ(araD-araB)567, ΔlacZ4787(::rrnB-3), λ-,rph-1, ΔdnaJ735::kan,Δ(rhaD-rhaB)568, hsdR514	(Baba *et al.*, 2006)	CGSC#: 8343
Δ*clpA* F-, Δ(araD-araB)567, ΔlacZ4787(::rrnB-3), λ-, ΔclpA783::kan,rph-1, Δ(rhaD-rhaB)568, hsdR514	(Baba *et al.*, 2006)	CGSC#: 8898
Δ*clpP* F-, Δ(araD-araB)567, ΔlacZ4787(::rrnB-3), ΔclpP723::kan, λ-,rph-1, Δ(rhaD-rhaB)568, hsdR514	(Baba *et al.*, 2006)	CGSC#: 8590
Δ*clpX* F-, Δ(araD-araB)567, ΔlacZ4787(::rrnB-3), ΔclpX724::kan, λ-, rph-1,Δ(rhaD-rhaB)568, hsdR514	(Baba *et al.*, 2006)	CGSC#: 8591
Δ*hslU* F-, Δ(araD-araB)567, ΔlacZ4787(::rrnB-3), λ-, rph-1, Δ(rhaD-rhaB)568, ΔhslU790::kan, hsdR514	(Baba *et al.*, 2006)	CGSC#: 10817

ΔhslV F-, Δ(araD-araB)567, ΔlacZ4787(::rrnB-3), λ-, rph-1, Δ(rhaD-rhaB)568, ΔhslV720::kan, hsdR514	(Baba *et al.*, 2006)	CGSC#: 10818
ΔsecB F-, Δ(araD-araB)567, ΔlacZ4787(::rrnB-3), λ-, ΔsecB721::kan, rph-1, Δ(rhaD-rhaB)568, hsdR514	(Baba *et al.*, 2006)	CGSC#: 10640
CSH4 F, lacZ1125, λtrpA49(Am), relA1, rpsL150(strR), spoT	(Newton *et al.*, 1965)	CGSC#: 8196
WG350 F-, lacZ1125, λ⁻, Δ(putA-putP)101, trpA49(Am), ΔproU-600, relA1, rpsL150(strR), spoT1, Δ(proP-melB)212	(Culham *et al.*, 1993)	CGSC#: 8195

Oligonucleotides

GFP Forward primer (to clone in pET SUMO) CCCGGATCCATGTCTAAAGGTGAAGAATTATTCACTGGTGTTGTCCC	This Paper	
GFP Reverse primer (to clone in pET SUMO) CCCAAGCTTTTATTTGTACAATTCATCCATACCATGGGTAATA	This Paper	
GFP Forward primer (to clone in pBAD vector) TGTGCTGAATTCACATATATGTCTAAAGGTGAAGAATTATTCA	This Paper	
GFP Reverse primer (to clone in pBAD vector) ACGGCCAAGCTTTTATTTGTACAATTCATCCATACCATGGG	This Paper	
RBS mCherry Forward primer (to clone in pBAD vector) ACAAAGCTTTCCATACCCGTTTTTTGGGCTAACAGGAGGAATTAACCATGGTGA	This Paper	
RBS mCherry Reverse primer (to clone in pBAD vector) TCTAAGCTTCTACTTGTACAGCTCGTC	This Paper	

Recombinant DNA

pOFX ELES (pOFX tac-SL2)	(Castanié *et al.*, 1997)[70]	
pOFX DnaK/J/GrpE (pOFX tac-KJE1)	(Castanié *et al.*, 1997)[70]	
pET duet1 dnaK	(Tiwari *et al.*, 2013)[71]	
pET duet1 dnaJ	(Tiwari *et al.*, 2013)[71]	
pET duet1 grpE	(Tiwari *et al.*, 2013)[71]	
pET duet1 Mge1	(Tiwari *et al.*, 2013)[71]	
pET SUMO sGFP	Sadat & Mapa (in process)	
pBAD Gm-R	(Bandyopadhyay *et al.*, 2012)[25]	
pBAD I26G Gm-R	This Paper	

pBAD I27G, R28G Gm-R	This Paper	
pBAD 37G Gm-R	This Paper	
pBAD 39G Gm-R	This Paper	
pET SUMO Wt GFP	This Paper	
pET SUMO C1GFP	This Paper	
pET SUMO C3FP	This Paper	
pET SUMO C6GFP	This Paper	
pBAD Wt GFP mCherry	This Paper	
pBAD C1GFP mCherry	This Paper	
pBAD C3GFP mCherry	This Paper	
pBAD C6GFP mCherry	This Paper	
Software and Algorithms		
Chimera		
Proteowizard	(Brown *et al.*, 2005)[72]	
XCMS	(Smith *et al.*, 2006; Silva *et al.*, 2014)[73,74]	
Maven	(Lu *et al.*, 2010; Clasquin, Melamud and Rabinowitz, 2012) [65,66]	
FastQC	http://www.citeulike.org/user/nailest/author/Andrews:S	
Trimmomatic	(Bolger, Lohse and Usadel, 2014)[61]	
Kallisto	(Zhang *et al.*, 2017; Bray *et al.*, 2016)[62,63]	
EBseq	(Leng *et al.*, 2013)[64]	
ImageJ		
Origin Pro8		
Deposited Data		

864
865

Figure 1: Schematic showing specific and general buffering mechanisms
Mutational buffering in a cell can be achived by modification in the regulatory region of a substrate leading to its increased expression, activation of alternate pathways or by increased expression of the binding partner for a specific substrate (top panel). However generalized buffering of mutant phenotypes is executed by protein folding assistance through molecular chaperones or through relative composition and concentration of metabolite and resultant changes in protein thermodynamics and kinetics (bottom panel).

Figure 2: Genetic alteration in cellular metabolism changes mutational buffering

A. Schematic of highthroughput activity assay for Glycine-doublet substitution (GG-mutant) library of Gm-R. Activity of Gm-R mutants is inferred from the competitive fitness of the mutants in presence of Gm-selection. The competitive fitness for each of the mutants is quantified by deep-sequencing.

B. Schematic of highthroughput activity assay for GFP mutant library. The bi-cistronic construct of GFP and mCherry is driven by an arabinose inducible promoter. RBS indicates the position of the additional Ribosome Binding Site for translation of mCherry to serve as internal control. Mutations are created on GFP using random mutagenesis with GFP specific primers. Pool of GFP mutants is sorted into population of compromised mutants (Pc) and active mutants (Pa) based on GFP Fluorescence.

C. Gene Ontology (GO) classes that were upregulated in WG350 transcriptome with respect to CSH4. The fold enrichment is shown on the left-axis and Benjamini-Hochberg FDR corrected p-values for the enrichment score are shown on the right axis.

D. Comparison of metabolic features measured in WG350 and CSH4 using untargeted metabolomics. Metabolite features that are significantly different between CSH4 and WG350 are represented as colored circles (p-value < 0.05, 5 biological replicates for each sample).

E. Normalized read counts of Gm-R GG-mutant library at comparative selective pressures in the presence of the antibiotic gentamicin (Gm). Pink shaded area marks the 99% confidence interval. Mutants marked in red show lower read-counts in WG350 than in CSH4.

F. Conservation score (calculated by consurf (Ashkenazy, H. *et al*, 2016)) for each residue. Higher consurf scores indicate more variable residues (less conserved among homologs. Each filled circle indicates the residues that are mutated in the clones studied.

G. Fractional ASA (calculated using VADAR (Willard, L. *et al*, 2003)) of the residues mutated in the clones is plotted against the temperature factor of the residues obtained from 1BO4 (left panel).

H. Distance of the residues mutated from the ligand was calculated from the pdb (1BO4) and plotted against the predicted alteration in protein stability. Positive ΔΔG corresponds to destabilization and negative to stabilization (calculated by foldX (Schymkowitz, J. *et al,* 2005)).

I. Left is a heat map representing growth of four Gm-R mutants in WG350 and in CSH4 along with Wt Gm-R in increasing concentration of Gm. Right panel shows the MIC of each of the mutants normalized with respect to Wt Gm-R in the respective strains. Error bars are representative of standard deviation from 4 biological replicates. Significance is calculated using Students' t-test with respect to CSH4 (* p-value < 0.05, *** p-value < 0.001).

Also see Figure S1

Figure 3: Osmotic stress-induced changes in metabolism changes the spectrum of mutants buffered

A. Gene Ontology (GO) classes that were upregulated in WG350 transcriptome with respect to CSH4 during osmotic shock. The fold enrichment is shown on the left-axis and Benjamini-Hochberg FDR corrected p-values for the enrichment score is shown on the right axis.

B. Comparison of metabolic features measured in WG350 and CSH4 during growth in media containing 350 mM NaCl using untargeted metabolomics. The significantly different metabolites (p-value < 0.05, 5 biological replicates for each sample) are represented in colored circles.

C. Comparison of metabolic features within strain in presence and absence of 350 mM NaCl in growth media. The significantly different metabolites (p-value < 0.05, 5 biological replicates for each sample) are represented in colored circles.

D. Normalized read counts of Gm-R GG-mutant library at comparative selective pressures in the presence of the antibiotic Gentamicin (Gm) and 350 mM NaCl in the growth media. Pink shaded area marks the 99% confidence interval.

E. Top panel shows heat map representing activity (in terms of growth) of Gm-R mutants described earlier in CSH4 and WG350 grown in presence of 350 mM NaCl and different concentration of Gm. Bottom panel shows the MIC of each of the mutants normalized with respect to Wt Gm-R in the respective strains. Error bars are representative of standard deviation from 4 biological replicates.

F. Single cell fluorescence for the partially active pool of GFP mutants (Pc). The scatter plot for mCherry Vs GFP/mCherry fluorescence is shown for Pc in WG350 and CSH4 when grown in the presence of 350 mM NaCl added in excess to the LB growth medium. Clones with high GFP fluorescence during osmotic stress in WG350 are indicated by the blue box.

G. Histogram of GFP/mCherry fluorescence of the Isolated clones (C1,C3,C6, and Wt GFP) in WG350 and CSH4 in the presence and absence of 350 mM NaCl. The shift in the ratio due to osmotic shock is shown with colored arrows. The cyan arrow indicates increase/decrease in GFP fluorescence upon addition of 350 mM NaCl to CSH4 strain, the orange arrow indicates the same for WG350.

H. Conservation score (calculated by consurf (Ashkenazy, H. *et al*, 2016)) for each residue. Higher consurf scores indicate more variable residues (less conserved among homologs. Each filled circle indicates the residues that are mutated in the clones studied.
I. Fractional ASA (calculated using VADAR (Willard, L. *et al*, 2003)) of the residues mutated in the clones is plotted against the temperature factor of the residues obtained from 1GFL (left panel). Positive ΔΔG corresponds to destabilization and negative to stabilization (calculated by foldX (Schymkowitz, J. *et al*, 2005)).

Also see Figure S2

Figure 4: Mutations buffered by altered metabolic states have compromised protein folding

A. Fluorescence of Wt GFP and GFP mutants (C1, C3 and C6) at 515 nm are shown at 200 nM concentration (excitation 488 nm /slit-width 2 nm, emission slit-width 5 nm).

B. Refolding rates of Wt and GFP mutants obtained on unfolding the proteins in 6 M GuHCl, followed by a 100-fold dilution into Buffer-A (50 mM Tris, 150 mM NaCl, 2 mM DTT, pH 7.4) to a final concentration of 200nm for the proteins. Refolding rates were obtained by fitting the refolding traces with single exponential equations.

C. Average melting temperature (Tm) calculated from thermal melts for Wt, C1, C3 and C6 GFP. Fluorescence of each mutants at different temperatures has been normalized to its respective fluorescence at $25^{O}C$. Standard deviations is plotted as error bars from 3 biological replicates.

D. Plot for Normalized GFP fluorescence of Wt, C1 and C6 GFP at 0 min, 30 min, 60 min and 120 min post translation arrest with Chloramphenicol after 5 hr induction of GFP with 0.5% arabinose in CSH4.

E. Plot for the gel-based quantification of intact GFP in Chloramphenicol-chase assay for monitoring degradation of GFP mutants and Wt GFP in CSH4.

F. Plot for the gel-based quantification of intact GFP in Chloramphenicol-chase assay for monitoring degradation of mutant GFPs C1 and C6 in CSH4, CSH4(S) (CSH4 grown in presence of 350 mM NaCl), WG350 and WG350(S) (WG350 grown in presence of 350 mM NaCl), Standard deviations is plotted as error bars from 3 biological replicates.

G. Plot for the gel-based quantification of intact GFP in Chloramphenicol-chase assay for monitoring degradation of sGFP in CSH4, CSH4(S), WG350 and WG350(S).

H. Schematic for simulation setting. Numerical simulation was set up with a fixed concentration of mRNA/ribosome complex (S). These complexes could either form nascent polypeptides (U) with a rate k_{trans} that is a collective rate constant for transcription/translation, or be inhibited with a rate constant k_{block} in the presence of a translation inhibitor (I). The pool of U could either degrade with a rate k_{deg} or fold with a rate constant of k_{f} and reach the native state (F). We finally monitor the total amount of undegraded protein (U+F) after blocking translation after 300 s of starting the simulation (simulation start mimics induction).

I. Total amount of intact GFP obtained from simulation as a function of k_{f} while keeping kdeg constant at 0.01 s^{-1}. Red line indicates the time-point at a 1 mM dose of translation-inhibitor (I) is added.

Also see Figure S3

Figure 5: Folding of isolated mutants is independent of molecular chaperones

A. Heatmap representing relative quantification of transcripts encoding proteostasis related proteins in WG350 and CSH4 in the presence and absence of 350 mM NaCl.

B. Left panel shows the immunoblot and right panel the image based quantitation of GroEL and DnaK expression levels in WG350 and CSH4 in the presence and absence of 350 mM NaCl (p-value > 0.05). Ponceau stained blot, used as loading control, is shown at the bottom of left panel.

C. Wt, C1 and C6 GFP were expressed in wild type *E. coli* (BW25113, BW) and in strains harboring single deletion of different molecular chaperones and proteases. GFP/mCherry for each of the mutants were normalized with respect to Wt GFP fluorescence in the same strains. Fold change in GFP/mCherry for each of the mutant GFPs between BW and the deletion strains was calculated using these normalized values. $\log_2$ values of this fold change is shown as heatmap.

D. Wt, C1 and C6 GFP were expressed in CSH4 or WG350 with and without plasmids overexpressing either GroEL/ES (top) or DnaK/J/E (bottom). Median GFP/mCherry fluorescence levels of different strains are shown as a bargraph. Error bars represent standard deviation of 3 biological replicates (p-value > 0.05).

E. Wt, C1 and C6 GFP were expressed in CSH4 and WG350 in the absence or presence of external stressors (42°C - heat shock, 80 μM Paraquat- oxidative stress, 350 mM NaCl - osmotic shock). GFP/mCherry fluorescence is shown in each of these conditions as bargraph with standard deviations as error bars from 3 biological replicates. Signifance is calculated using Students' t-test (* p-value < 0.05, ** p-value < 0.01).

F. To check chaperone dependence of GFP mutants *in vitro*, unfolded proteins were diluted 100-fold in the presence of DnaK/DnaJ/GrpE (400 nM/800 nM/400 nM, respectively) or GroEL/ES (400 nM/800 nM, respectively), and folding was initiated by adding 2 mM ATP. Refolding was followed by monitoring GFP fluorescence as a function of time. Rate of refolding is plotted for C1 and C6 mutants of GFP in presence and absence of chaperoning machinery with error bars from 3 biological replicates.

G. Immunoblotting for GFP showing the recruitment of nascently formed C1 GFP into multiprotein complexes. After a brief induction of GFP expression, cells were treated with a cell-permeable crosslinker (DSG). Subsequent to crosslinking, cells were lysed and resolved on SDS-PAGE followed by immunoblotting with anti-GFP antibody. Free (or monomeric) GFP is shown with blue box, dimeric GFP is shown in red box and multimeric complexes are shown in black box. The ratio of GFP in multimeric complex to the amount in monomeric state is shown in the accompanying bar graph.

H. Histogram of GFP/mCherry fluorescence of Wt GFP and pool of GFP mutants having compromised fluorescence (Pc) in BW, molecular chaperone (Δ*dnaJ,* Δ*dnaK,* Δ*tig*) and protease knockout strains (Δ*lon*).

I. Plot for the gel-based quantification of intact GFP in Chloramphenicol-chase assay for monitoring degradation of Wt and C1 GFP in BW, chaperone (Δ*dnaK*), and protease (Δ*lon*) knockout strain. Standard deviation is plotted as error bars from 3 biological replicates.

Also see Figure S4

Figure 6: Metabolites can chaperone the mutants *in vivo* and *in vitro*.
A. Refolding of Wt GFP or mutant GFP was initiated as described earlier in the presence of 100 mM concentration of different metabolites. Refolding rates were obtained by fitting the refolding traces with single exponential equations. Mean of refolding rates, along with standard deviations are shown for three replicates of refolding reactions. RB (shown in green bars) are refolding rates in Buffer-A, the base buffer in which the metabolites are added. p-value is calculated using Students' t-test with repect to RB (* p-value < 0.05, ** p-value < 0.01, *** p-value < 0.001).
Ala: Alanine, Asp: Aspartate, βala: β alanine, Bet: Betaine, FBP: Fructose-1,6-bisphosphate, Fum: Fumarate, Gluc: Glucose, Gly: Glycine, Mal: Malate, Malo: Malonate, Pro: Proline, Ser: Serine, Sor: Sorbitol, Suc: Succinate, Sucr: Sucrose, Tre: Trehalose.
B. Protein free extracts were isolated from wild type *E. coli* (BW25113, BW) as described in Methods. Refolding was performed by diluting unfolded proteins (Wt GFP or GFP mutants) 100-fold into these extracts.

Figure 7: Altering cellular metabolism by exogenous supplementation of metabolites modifies mutational buffering capacity

A. Expression of Wt, C1 and C6 GFP mutants was induced in CSH4 cells either growing in LB or in LB containing different metabolites. Log_2 of fold change in median of GFP/mCherry fluorescence as measured by single-cell fluorescence is shown as heat map

B. Fold change in median of GFP/mCherry fluorescence in few metabolites (Alanine, Betaine, Proline, Aspartate) in CSH4 is shown as bar graph along with standard deviations from three biological replicates and P-value caluclated with respect to Wt GFP by Students' t-test, *p-value <0.05.

Figure 8: Wild-type cells evolve altered buffering capacity with evolution of a different metabolome

A. Schematic of strategy for Laboratory Adaptive Evolution of osmotolerant strains of *E. coli* starting from BW (wild type *E. coli* K-12, BW25113).

B. Growth curve of unevolved BW and evolved osmotolerant strains E5, F12 and H8 in 500 mM of NaCl added in excess to LB medium while growing at 37°C, 200 rpm.

C. Genetic interaction network map of evolved strains E5 and H8 based on mutations obtained in genome sequencing.

D. Metabolite features of representative evolved osmotolerant strains (E5, F12 and H8) and its comparison with BW. The colored circled represent metabolites that are significantly altered in the evolved strains.

E. Histogram for GFP/mCherry fluorescence of Wt, C1 and C6 GFP in the BW and evolved strains (E5, F12 and H8).

F. Comparative quantification of chaperone proteins DnaK and GroEL in BW and evolved strains (E5, F12 and H8) done by immuno-blotting with specific antibodies as described earlier. Quantification of the chaperone levels are shown as bar graphs. Error bars represent standard deviation of 3 replicate measurement (p-value > 0.05).

G. Chloramphenicol based chase for checking protein degradation rates were performed in BW and evolved strains (E5, F12 and H8) as discussed earlier. The levels of C1 GFP at different time points are shown as a function of time in different strains.

H. Heatmap for growth based activity of Gm-R mutants and Wt Gm-R in unevolved strain (BW) and evolved strains (E5, F12 and H8) in increasing concentrations of gentamicin.

I. Bar graph representing percentage MIC of a selected set of Gm-R mutants w.r.t Wt Gm-R in the respective strains. Error bars represent standard deviation from 2 biological replicates.

Also see Figure S5.

FIGURE S1: Strain specific metabolic differences lead to differential mutation buffering capacity (Related to Figure 2)

A. Schematic of Glycine doublet substitution library (GG) of Gm-R mutants where two consecutive amino acids have been substituted with two Glycine residues (red solid filled) starting from 5th amino acid.

B. Growth curve for *E.coli* strains CSH4 (black, blue) and WG350 (red, green) in presence and absence of 350 mM NaCl added in excess to the LB media while growing at 37°C.

C. Scatter plot for log$_2$ of fold change in metabolite concentrations between CSH4 and WG350 against -log$_{10}$ of p-value. The significantly altered metabolites (p-value < 0.05, 5 biological replicates for each sample) are shown in colored circles

D. Pathway for production of glycine betaine in *E.coli* showing correlation between metabolomics and transcriptomics. Metabolites upregulated are shown as circles and transcripts for the genes are shown as squares. Color intensity corresponds to fold upregulation

E. Growth of untransformed CSH4 (-), WG350 (-) and cells transformed with Wt Gm-R and Glycine doublet mutant library plasmids (M-Lib) in increasing concentrations of Gentamicin (0-800 mg/ml). Increase in growth is shown as increase in color density.

F. Correlation between NGS based quantification of enrichment scores (z-score) and MIC based semi-quantitative activity measurement for Wt Gm-R and mutants I26G, I27G- R28G, V37G-K38G, S39G-K40G of Glycine doublet mutant library.

G. Histogram for *in vivo* fluorescence of Wt GFP (represented as ratio of GFP/mCherry) in CSH4 (black) and WG350 (red)

H. Histogram for ratio of GFP/mCherry fluorescence (left) and independent fluorescence of GFP (middle) and mCherry channel (right) of mutant library Pc in CSH4 and WG350.

FIGURE S2: Osmotic stress induced modification in cellular metabolism changes the subset of mutations buffered (Related to Figure 3)

A. Scatter plot for $\log_2$ of fold change in metabolic features between WG350 and CSH4 during osmotic shock against p-value. The significantly different metabolites (p-value < 0.05, 5 biological replicates for each sample) are represented in colored circles.

B. Scatter plot for $\log_2$ of fold change in metabolic features within strains CSH4 (left) and WG350 (right) on osmotic shock(s) against p-value. The significantly different metabolites (p-value < 0.05, 5 biological replicates for each sample) are represented in colored circles.

C. Amino acid sequence alignment of Wt GFP (YeGFP) with the isolated mutants C1, C3 and C6. Mutations are shown in red boxes.

D. Histogram for GFP (top panel) and mCherry fluorescence (bottom panel) of the Isolated clones (Wt, C1, C3 and C6 GFP) in WG350 and CSH4 in the presence and absence of 350 mM NaCl.

FIGURE S3: Fluorescence increase is not due better proteostasis in WG350(S) (Related to Figure 4)

A. Mutations buffered in WG350 while growing in 350 mM NaCl (C1, C3, and C6) are marked on the GFP crystal structure (pdb: 1GFL) (Yang, Moss and Phillips, 1996). Fluorophore at the centre of the barrel is highlighted in magenta.

B. Refolding traces of Wt and GFP mutants obtained on unfolding the proteins in 6 M GuHCl, followed by a 100-fold dilution into Buffer-A (50 mM Tris, 150 mM NaCl, 2 mM DTT, pH 7.4) to a final concentration of 200 nm for the proteins. The GFP fluorescence (AU) is plotted against time (s).

C. Average thermal melt curves for Wt, C1, C3 and C6 GFP. Fluorescence of each mutants at different temperatures has been normalized to its respective fluorescence at 25°C.

D. Schematic of Chloramphenicol-based chase for degradation of GFP as described in star methods. After a short (5 min) induction of GFP with 0.5% arabinose, translation was stalled with Chloramphenicol. Cells were lysed at different points and the amount of intact GFP was monitored by immunoblotting with anti-GFP antibody.

E. Plot for the gel-based quantification of intact GFP (left) and representative images for immunoblotting to chase degradation of Wt, C1 and C6 GFP in CSH4 after 5 hr induction of GFP with 0.5% arabinose followed by translation arrest with Chloramphenicol. Ponceau has been used as loading control.

F. Representative images for immunoblotting to chase degradation of nascent chains of C1 and C6 GFP as described in Star Methods in CSH4 and WG350 in presence and absence of osmotic stress. GFP degrades slowest in WG350(S). Ponceau has been used as loading control.

G. Representative images for immunoblotting to chase degradation of sGFP in CSH4 and WG350 in presence and absence of osmotic stress. Ponceau has been used as loading control.

Figure S4: Observed buffering of GFP mutants is not channeled only through molecular chaperones (Related to Figure 5)

A. Histogram for GFP/mCherry Fluorescence (top panel), GFP (middle panel) and mCherry fluorescence (bottom panel) in absence of overexpressed GroEL/ES (black, red, 2 biological replicates) and while co-expressed with GroEL/ES (blue, green, 2 biological replicates) in CSH4 and WG350. Induction protocol was followed as described in Star Methods.

B. Histogram for GFP/mCherry Fluorescence (top panel), GFP (middle panel) and mCherry fluorescence (bottom panel) in absence of overexpressed DnaK/J/GrpE (black, red, 2 biological replicates) and while co-expressed with DnaK/J/GrpE (blue, green, 2 biological replicates) in CSH4 and WG350.

C. Representative images for immunoblotting to chase degradation of Wt and C1 GFP in WT *E.coli* K12 strain BW25113 (BW) chaperone (Δ*dnaK*), and protease (Δ*lon*) knockout strain. Ponceau has been used as loading control.

D. Immunoblotting for GFP showing the recruitment of nascently formed C6 GFP into multiprotein complexes. Free (or monomeric) GFP are shown with blue box, dimeric GFP are shown in red box and the multimeric complexes are shown in black box. The ratio of GFP in multimeric complex and the amount in monomeric state is shown in the accompanying bar graph.

E. Histogram of GFP/mCherry fluorescence of Wt GFP and pool of GFP mutants having compromised fluorescence (Pc) in BW and molecular chaperone knockout strains (Δ*dnaJ*, Δ*dnaK*, Δ*tig*, Δ*lon*).

I. Representative images for immunoblotting of chase degradation of Wt and C1 GFP in BW, chaperone (Δ*dnaK*), and protease (Δ*lon*) knockout strain. Ponceau has been used as loading control.

G. Heatmap for growth based activity of Wt and Gm-R mutant library (lib1, lib2) (top panel), Wt Gm-R anf Gm-R mutants (I26G, I27G-R28G) in BW and molecular chaperone knockout strain (Δ*dnaK*) in increasing concentrations of gentamicin.

Figure S5: Different small molecules affect energetics of folding in a mutant specific manner (Related to Figure6)
A. Thermal melting profile of Wt, C1, C3 and C6 GFP as change in melting temperature (ΔTm) in presence of
different small molecules. The change in Tm in presence of a small molecule is obtained by subtracting Tm in presence
of Buffer-A (50 mM Tris, 150 mM NaCl, 2 mM DTT, pH 7.4) from Tm in presence of that respective small molecule (100 mM).
Error bars indicate standard deviation among 3 technical replicates.
Suc: Succinate, Sucr: Sucrose, Ala: Alanine, Bet: Betaine, Arg: Arginine, Pro: Proline, Mal: Malate, Malo: Malonate, Ser: Serine,
Sor: Sorbitol, Fum: Fumarate, Asp: Aspartate, TMAO: Trimethylamine-N-oxide

Figure S6: Modification in metabolome due to evolution changes the mutation buffering capacity of the cell (Related to Figure 8)

A. Histogram for fluorescence of Wt and GFP mutants C1 and C6 represented as GFP (left panel) and mCherry fluorescence (right panel) in unevolved BW (WT E.coli K 12, BW25113) and evolved osmotolerant strains E5, F12, H8.

B. Representative images of Chloramphenicol based chase for checking degradation rates of C1 GFP in BW (WT *E.coli* K 12, BW25113) and evolved strains E5, F12, H8. GFP is degraded slower in evolved strains than BW

C. Representative images of Chloramphenicol based chase for sGFP in BW (WT *E.coli* K 12, BW25113) and evolved strains E5, F12, H8. sGFP is degraded with similar rates in evolved and unevolved BW as shown in densitometric analysis plot at the bottom

Table S1: List of metabolites along with their log$_2$ of fold change and p- values obtained by untargeted metabolite profiling and analysis using MAVEN platform (Related to Figure 2 and 3)

S.no.	med Mz	med Rt	Max Quality	compound	Log$_2$ fold change				Log$_{10}$ p-value			
					CSH4 - WG350	CSH4 - CSH4 (S)	WG350 - WG350 (S)	CSH4 (S)-WG350 (S)	CSH4-WG350	CSH4-CSH4 (S)	WG350 - WG350 (S)	CSH4 (S)-WG350 (S)
1	363.0344	2.041682	0.848391	N-Acetylputrescine	0.00	-14.03	-14.29	-0.26	0.01	0.97	1.54	0.45
2	283.0684	4.029608	0.849596	Geranyl-PP	0.00	-13.99	-13.89	0.10	0.05	1.67	1.43	0.20
3	151.025	1.858412	0.819399	fructose-1-6-bisphosphate	0.00	-12.65	0.00	12.65	0.00	2.24	0.43	5.16
4	116.0705	2.341379	0.828605	succinate	0.00	-11.67	-25.78	-14.11	0.43	1.60	2.15	1.80
5	243.0622	2.131804	0.848934	UMP	2.10	-4.63	-4.86	1.87	0.49	2.56	1.80	1.80
6	167.0204	2.374183	0.847362	orotidine-5--phosphate	10.98	-4.25	-14.62	0.61	1.12	4.78	1.71	1.38
7	323.0291	3.537884	0.851401	betaine	-0.93	-4.17	-7.94	-4.70	0.53	0.95	0.98	1.78
8	606.0748	4.036123	0.851434	Octoluse Bisphosphate	1.06	-3.88	-4.69	0.24	0.15	1.00	1.04	0.28
9	565.0482	4.056136	0.857477	tryptophan	0.44	-3.75	-4.18	0.01	0.05	1.93	1.71	0.22
10	402.99499	2.973083	0.84958	thiamine-phosphate	0.12	-3.73	-2.99	0.87	0.01	1.34	1.15	3.48
11	180.0657	2.219842	0.858241	fumarate	-2.93	-3.27	-6.82	-6.49	0.88	0.51	3.44	1.03
12	203.0819	1.413762	0.840261	FMN	1.54	-2.86	-1.06	3.34	1.17	1.98	0.65	4.25
13	203.0819	1.505575	0.858171	asparagine	-1.04	-2.73	0.10	1.79	0.62	2.12	0.10	3.83
14	341.1089	2.038149	0.846411	glutathione disulfide	-1.96	-2.52	-0.13	0.43	1.29	1.15	0.39	0.24
15	421.0756	1.887725	0.851835	CMP	-1.08	-2.37	1.97	3.26	0.47	1.60	1.10	1.82
16	241.0827	2.019653	0.856852	GMP	1.22	-2.33	-2.11	1.44	1.97	4.11	2.86	3.62
17	344.0691	2.006949	0.846896	UDP	-0.13	-2.28	-1.68	0.47	0.10	4.33	3.04	0.73
18	424.039	1.499295	0.860667	aspartate	-0.40	-1.99	2.16	3.75	0.64	3.32	2.82	5.07
19	264.1045	1.446013	0.858174	trehalose/sucrose	1.55	-1.89	-1.60	1.84	3.11	3.67	2.71	3.08

20	866.1249	1.45047	0.857708	Cellobiose	1.55	-1.89	-1.60	1.84	3.11	3.67	2.71	3.08
21	117.0186	2.093047	0.854168	N-Acetyl-L-alanine	-1.80	-1.77	0.08	0.05	0.78	1.01	0.38	0.06
22	171.0065	3.843684	0.849186	IMP	-0.41	-1.66	-1.07	0.18	0.38	2.09	1.54	0.45
23	253.0099	2.067247	0.861783	Pyroglutamic acid	-1.85	-1.53	-0.12	-0.43	0.64	0.67	0.41	0.80
24	368.9989	3.985863	0.849202	S-methyl-5--thioadenosine	-0.27	-1.28	-1.28	-0.28	0.15	1.34	1.39	0.43
25	266.0707	1.685908	0.833351	ribose-phosphate	-0.60	-1.15	0.82	1.37	0.75	1.65	1.10	2.80
26	296.0806	2.066397	0.852619	N-acetyl-L-ornithine	0.30	-1.10	-1.11	0.30	0.35	2.75	2.67	0.64
27	383.1152	3.042196	0.771733	Acetyllysine	0.05	-0.96	-0.98	0.04	0.11	1.80	2.12	0.13
28	229.0112	2.020989	0.85325	adenosine	0.89	-0.91	-1.32	0.48	0.93	2.14	2.15	2.26
29	375.131	2.323842	0.84781	xanthosine-5-phosphate	-0.18	-0.89	-0.73	-0.01	0.04	1.82	2.08	0.03
30	176.935	4.759889	0.852717	2-Isopropylmalic acid	0.23	-0.74	-0.66	0.30	0.24	1.32	1.66	0.68
31	128.0341	1.83335	0.853449	dCDP	-0.36	-0.47	0.46	0.58	0.07	1.29	0.49	2.60
32	168.0566	2.160493	0.853901	xanthosine	0.45	-0.45	-0.52	0.37	1.07	1.80	1.32	1.07
33	114.0551	10.95572	0.73141	p-hydroxybenzoate	-0.33	-0.42	-0.22	-0.14	0.67	0.25	1.34	0.49
34	225.0392	11.50664	0.84889	Sedoheptoluse bisphosphate	-0.26	-0.24	-0.07	-0.09	0.28	0.00	0.14	0.08
35	145.0301	1.69376	0.833608	cystathionine	0.19	-0.23	-1.06	-0.64	0.54	0.38	1.94	0.55
36	218.1029	1.981234	0.860063	hypoxanthine	-0.22	-0.19	-0.01	-0.04	0.52	0.74	0.11	0.03
37	137.0232	1.572158	0.857927	glutamine	-0.07	-0.16	-0.09	0.00	0.28	0.01	0.07	0.06
38	367.0198	1.393888	0.84899	UDP-D-glucose	-0.14	-0.14	-0.10	-0.09	0.64	0.59	0.14	0.02
39	131.0814	6.98723	0.856877	Kynurenine	0.22	-0.11	-0.34	-0.02	0.22	0.36	1.13	0.16
40	399.0131	5.573004	0.850927	riboflavin	-0.10	-0.09	-0.13	-0.14	0.07	0.04	0.50	0.29
41	319.0463	5.199954	0.849735	Thiamine pyrophosphate	-0.03	-0.09	-0.05	0.01	0.12	0.54	0.56	0.12

42	146.0447	5.07694	0.840613	Octoluse 8/1P	0.10	-0.08	-0.32	-0.14	0.20	0.50	0.53	0.25
43	742.0695	3.301201	0.735053	valine	0.22	-0.08	-0.36	-0.06	0.27	0.36	1.11	0.22
44	664.118	1.939427	0.662212	cyclic-AMP	0.27	-0.02	0.62	0.91	0.55	0.64	1.54	1.10
45	662.1027	8.18236	0.852802	tryptophan	0.28	-0.01	-0.40	-0.11	0.20	0.13	0.99	0.61
46	175.0351	1.743987	0.85161	glucose-6-phosphate	0.35	0.04	1.96	2.27	0.75	0.07	0.43	3.75
47	129.1023	6.07219	0.850308	thiamine	0.00	0.05	-0.08	-0.14	0.13	0.05	0.50	0.34
48	173.0922	2.708844	0.81459	pantothenate	0.40	0.06	-0.53	-0.18	0.33	0.03	1.38	0.97
49	130.0498	5.998872	0.854558	biotin	0.14	0.06	-0.17	-0.09	0.19	0.08	0.77	0.16
50	187.0716	1.777959	0.832413	shikimate-3-phosphate	0.04	0.06	0.20	0.18	0.44	0.47	1.26	0.37
51	188.0556	1.504867	0.772704	N-acetyl-glutamine	0.04	0.09	0.16	0.10	0.02	0.01	1.50	1.00
52	300.0466	1.719046	0.848407	guanine	-0.13	0.13	0.24	-0.02	0.45	0.03	0.79	0.22
53	179.0552	1.501844	0.85144	inosine	-0.21	0.17	1.76	1.37	0.00	0.07	2.49	1.18
54	134.0294	2.00565	0.855251	dGMP	0.40	0.25	0.12	0.27	0.68	0.33	0.71	0.24
55	148.0426	3.28745	0.850251	D-sedoheptulose-1/7-phosphate	0.19	0.26	0.12	0.05	0.03	0.17	0.99	0.45
56	133.0131	1.936306	0.677158	methionine	-0.17	0.29	0.50	0.05	0.16	1.54	0.85	0.13
57	145.0971	1.390545	0.850053	UDP-N-acetyl-glucosamine	-0.21	0.40	0.18	-0.42	0.17	0.40	0.66	0.17
58	205.0347	2.536442	0.851966	proline	-0.88	0.40	-1.20	-2.48	0.33	0.24	0.81	1.84
59	289.1155	1.833144	0.851898	a-ketoglutarate	0.38	0.42	1.06	1.02	1.23	1.41	2.15	1.78
60	207.077	2.012048	0.833829	tyrosine	0.15	0.48	0.51	0.18	0.70	2.96	2.09	1.13
61	267.0725	1.519962	0.766428	Aminoadipic acid	0.22	0.49	0.01	-0.26	0.38	1.51	0.22	1.52
62	347.0395	2.04751	0.799432	N-acetyl-glutamate	0.26	0.53	0.47	0.19	0.21	1.07	0.92	0.36
63	135.0301	3.948531	0.850648	acetyl-CoA	0.12	0.55	1.77	1.34	0.04	2.12	2.55	2.32
64	112.4286	1.409344	0.858314	arginine	0.12	0.56	0.53	0.09	0.51	1.82	2.03	0.09

65	154.0611	1.914433	0.856739	glutathione	0.05	0.57	0.97	0.45	0.03	1.40	2.23	1.89
66	282.0861	1.459852	0.857308	ornithine	-0.04	0.58	0.55	-0.07	0.34	0.41	2.33	0.09
67	150.0411	2.037107	0.853763	thymidine	0.18	0.59	0.36	-0.04	0.44	1.94	2.04	0.31
8	362.0511	1.499073	0.851382	L-arginino-succinate	-0.38	0.63	2.28	1.27	0.38	1.49	2.82	0.84
69	611.1408	2.658341	0.853422	N-acetyl-glucosamine-1/6-phosphate	0.02	0.67	0.51	-0.14	0.21	2.95	2.74	0.61
70	306.0765	1.728165	0.823201	N-carbamoyl-L-aspartate	0.61	0.73	0.23	0.11	0.87	0.40	0.17	0.02
71	145.0607	2.06878	0.715853	uridine	0.12	0.75	0.56	-0.06	0.21	2.06	1.65	0.42
72	259.022	1.509566	0.783648	pyridoxine	-0.08	0.76	0.58	-0.26	0.11	3.64	3.39	2.73
73	178.0714	1.853829	0.854485	sn-glycerol-3-phosphate	0.08	0.83	0.36	-0.39	0.00	2.49	1.50	1.29
74	177.0395	2.04771	0.750985	xanthine	-0.02	0.84	0.62	-0.24	0.23	2.54	2.09	1.04
75	313.0589	2.024257	0.820586	citrate/isocitrate	0.48	0.85	0.79	0.42	0.68	1.42	1.53	1.24
76	109.5264	1.386311	0.855701	adenosine 5--phosphosulfate	0.31	0.85	0.50	-0.04	0.53	2.48	2.30	0.33
77	338.989	1.485925	0.818313	O-acetyl-L-serine	-0.14	0.93	0.85	-0.22	0.73	3.92	3.35	1.01
78	455.1014	6.339029	0.859511	2_3-dihydroxybenzoic acid	-1.45	0.94	-0.60	-2.99	2.03	1.58	1.11	3.67
79	784.1508	1.505536	0.850999	cytidine	-0.05	0.96	0.86	-0.15	0.01	4.59	3.03	0.30
80	321.0496	1.780281	0.860946	malate	-0.10	0.99	0.15	-0.95	0.12	1.90	0.05	4.07
81	346.0558	1.441355	0.856271	lysine	1.42	1.02	-0.06	0.34	1.83	1.76	0.14	1.05
82	686.1423	1.567711	0.853065	lipoate	0.04	1.08	0.61	-0.43	0.38	3.09	2.51	2.90
83	251.0776	4.043802	0.84958	FAD	-0.14	1.11	0.56	-0.69	0.23	3.17	0.87	1.06
84	250.0937	2.100759	0.846352	NADP+	0.30	1.11	1.08	0.27	0.04	0.65	1.58	0.25
85	386.0172	1.486119	0.856649	D-gluconate	-0.02	1.15	1.23	0.05	0.30	2.63	3.13	0.20
86	289.0352	1.507047	0.851621	2-dehydro-D-gluconate	-0.21	1.28	1.75	0.27	0.04	2.48	1.42	0.27

87	258.0375	1.453563	0.828056	histidine	0.30	1.30	1.00	0.01	0.46	2.73	3.55	0.11
88	195.0502	2.024109	0.852796	dTMP	-0.94	1.31	1.59	-0.66	1.19	1.38	2.20	1.03
89	199.0007	1.550444	0.856554	homoserine	-2.41	1.32	-1.46	-5.18	0.76	0.25	0.97	1.45
90	242.078	2.027355	0.788688	glucono-Î´-lactone	0.47	1.33	1.05	0.18	0.65	2.37	1.75	0.20
91	239.0147	1.507822	0.829343	myo-inositol	-0.17	1.38	1.11	-0.44	0.91	4.50	3.63	1.23
92	221.0598	1.489268	0.854986	citrulline	-0.20	1.46	1.22	-0.43	0.35	3.66	2.77	1.07
93	328.0454	3.954113	0.838432	succinyl-CoA/methylmalonyl-CoA	-0.26	1.66	1.51	-0.42	0.34	1.03	2.36	0.81
94	689.0877	2.038819	0.852447	cyclic bis (3-->5-) dimeric GMP	-0.18	1.68	1.61	-0.24	2.83	3.02	2.29	0.14
95	766.1091	1.525817	0.835367	Phenylpropiolic acid	0.14	1.75	1.23	-0.37	0.00	0.58	1.85	0.47
96	322.0445	3.260484	0.848054	coenzyme A	-1.70	1.77	1.48	-1.99	1.48	2.14	1.76	3.87
97	174.0874	1.793035	0.851756	D-erythrose-4-phosphate	0.19	1.89	1.83	0.14	0.42	1.66	1.94	0.01
98	191.019	1.979981	0.851608	Methylcysteine	0.95	2.07	1.70	0.58	0.58	1.09	0.71	0.54
99	341.1089	1.499742	0.852927	S-ribosyl-L-homocysteine-nega	0.12	2.14	0.40	-1.63	0.11	2.81	0.72	1.16
100	243.0804	1.504422	0.852072	deoxyinosine	0.31	2.21	1.33	-0.57	0.79	2.89	2.21	0.13
101	116.0711	1.985581	0.85367	NADH	-2.59	2.54	3.62	-1.52	1.44	0.68	3.51	0.63
102	132.0291	1.759809	0.850709	trehalose-6-Phosphate	2.73	2.56	0.30	0.47	3.62	3.26	0.95	2.36
103	131.045	1.46988	0.853972	deoxyadenosine	0.33	2.69	1.93	-0.43	0.47	4.44	2.27	0.21
104	173.1035	1.516454	0.850518	D-glucosamine-6-phosphate	0.68	2.77	1.61	-0.49	0.69	1.86	0.67	0.30
105	160.0605	1.466613	0.852018	glucosamine	0.42	2.83	2.92	0.51	0.89	3.49	2.96	0.18
106	426.0121	2.080937	0.848104	NAD+	-0.69	3.58	2.11	-2.15	0.07	2.46	0.79	1.78
107	266.091	1.819532	0.852153	Pyrophosphate	0.22	4.02	3.37	-0.43	0.55	5.90	4.13	0.82
108	187.108	1.889188	0.850813	Uric acid	-0.31	4.32	3.55	-1.08	0.17	4.38	2.87	0.74
109	808.1196	2.336574	0.830013	dephospho-CoA	-1.36	11.89	2.10	-11.15	1.16	1.83	1.11	0.44

110	145.0132	2.367599	0.838086	Cystine	1.04	12.24	11.20	0.00	2.09	7.53	1.86	0.43
111	153.0182	1.50954	0.803857	S-adenosyl L- homoCyst eine	-0.86	12.98	13.85	0.00	1.67	8.01	6.35	#DIV/0!
112	175.0602	1.598831	0.850469	prephenat e	0.46	13.61	13.14	0.00	0.07	5.34	1.52	0.75
113	193.0346	1.51109	0.831777	guanosine	0.08	14.42	1.34	-13.00	0.45	1.36	2.89	1.14

Key: med Mz: Median *m/z* for metabolites, medRt: Median Retention Time, Max Quality: quality of peaks detected, compound: Metabolite, CSH4 - WG350: $\log_2$ of fold change and between CSH4 and WG350, CSH4 - CSH4 (S): $\log_2$ of fold change between CSH4 and CSH4 salt, WG350 - WG350 (S): $\log_2$ of fold change between WG350 and WG350 salt, CSH4 (S)- WG350 (S):): $\log_2$ of fold change between CSH4 salt and WG350 salt

Table S2: ODEs used for the kinetic simulations. (Related to figure 4)

Numerical simulation was set up with a fixed concentration of mRNA/ribosome complex (S). These complexes could either form nascent polypeptides (U) with a rate ktrans that is a collective rate constant for transcription/translation, or be inhibited with a rate constant kblock in the presence of a translation inhibitor (I). The pool of U could either degrade with a rate kdeg or fold with a rate constant of kf and reach the native state (F). We finally monitor the total amount of undegraded protein (U+F) after blocking translation after 300 s of starting the simulation (simulation start mimics induction). The reaction volume was kept at 1 liter.

ODEs:

d(U)/dt = -ReactionFlux1 - ReactionFlux2 + ReactionFlux3

d(F)/dt = ReactionFlux1

d(S)/dt = -ReactionFlux3 - ReactionFlux4

d(B)/dt = ReactionFlux4

d(I)/dt = -ReactionFlux4

Fluxes:

ReactionFlux1 = kf*U

ReactionFlux2 = kd*U

ReactionFlux3 = ktrans*S

ReactionFlux4 = ki*S*I

Parameter Values:

kf = 0.02 (s-1)

kdeg = 0.01 (s-1)

ktrans = 0.001(s-1)

ki = 0.0001 (M-1s-1)

cell volume = 1 (liters)

Initial Conditions (micromoles):

U = 0

F = 0

S= 1000

B = 0

I = 0

Dosing:

Delay: 300 s

1 mM of I added to reaction

Table S3: Concentration of metabolites added to LB to measure effect of metabolites on Fluorescence *in vivo* (Related to Figure 7)

S. no.	Metabolite	Concentration (mM)
1	Alanine	200
2	Arginine	2.5
3	Betaine	200
4	Proline	200
5	Serine	200
6	TMAO	200
7	Aspartate	200
8	Malate	100
9	Malonate	100
10	Fumarate	100
11	Succinate	100
12	KCL	350
13	Sucrose	400
14	Sorbitol	400

Table S4: Identity of genes along with the type of mutation observed in the evolved thermotolerant strains E5 and H8 on genome sequencing (Related to Figure 8)

Strain	Gene	Type
H8	mngA	Ins del
H8	hyfB	Ins del
H8	fucI	Ins del
H8	kefC	missense
H8	sgrR	missense
H8	mraY	missense
H8	ftsW	missense
H8	dxs	missense
H8	cstA	stop
H8	mngA	missense
H8	clpA	missense
H8	flgL	missense
H8	oppB	missense
H8	uidR	missense
H8	otsA	missense
E5	mngA	Ins del
E5	csiR	Ins del
E5	fucI	Ins del
E5	rihC	missense
E5	lptD	missense
E5	mraY	missense
E5	malZ	stop
E5	cyoA	missense
E5	paaH	missense
E5	sdaA	missense