EfgA is a conserved formaldehyde sensor that halts bacterial translation in response to elevated formaldehyde

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Abstract

Normal cellular processes give rise to toxic metabolites that cells must mitigate. Formaldehyde is a universal stressor and potent metabolic toxin that is generated in organisms from bacteria to humans. Methylotrophic bacteria such as *Methylorubrum extorquens* face an acute challenge due to their production of formaldehyde as an obligate central intermediate of single-carbon metabolism. Mechanisms to sense and respond to formaldehyde were speculated to exist in methylotrophs for decades but had never been discovered. Here we identify a member of the DUF336 domain family, named *efgA* for enhanced formaldehyde growth, that plays an important role in endogenous formaldehyde stress response in *M. extorquens* PA1 and is found almost exclusively in methylotrophic taxa. Our experimental analyses reveal that EfgA is a formaldehyde sensor that inhibits translation in response to elevated levels of formaldehyde. Heterologous expression of EfgA in *Escherichia coli* increases formaldehyde resistance, indicating that its interaction partners are widespread and conserved and may include translational machinery. EfgA represents the first example of a formaldehyde stress response system that does not involve enzymatic detoxification. Thus, EfgA comprises a unique stress response mechanism in bacteria, whereby a single protein directly senses elevated levels of a toxic intracellular metabolite and modulates translational activity.
Introduction

Robust organisms employ various mechanisms for averting cellular damage in the face of stress. In all biological systems, routine cellular processes generate highly toxic metabolites that can inflict damage on macromolecules and metabolites. The response systems that mitigate these endogenous stressors vary from detoxification systems, to neutralize reactive compounds such as hydrogen peroxide, imines, and aldehydes [1–4], to post-damage repair systems such as enzyme-mediated damage reversal [5,6] and targeted molecular degradation systems [7]. Induction of these responses can arise due to direct sensing of the metabolic toxins [8] or by sensing the damaged molecules themselves [5].

Formaldehyde is a ubiquitous metabolic toxin generated in most, if not all, organisms as a byproduct of enzymatic reactions or degradation products of metabolites. Due to its high reactivity with amines and thiols in particular, formaldehyde can damage numerous molecules such as metabolites [9], nucleic acids [10], and proteins [11–13]. To date, the only known formaldehyde-specific stress response systems involve enzymatic detoxification [2,14–16]. In bacteria, various formaldehyde detoxification pathways exist, including the widely conserved glutathione- (GSH-) dependent pathway as well as pathways dependent on pterins or sugar phosphates [2]; some species employ multiple pathways [17]. Thus far there is a single example of a formaldehyde sensor, FrmR, a transcriptional repressor that directly binds formaldehyde and controls expression of the detoxification pathway in *Escherichia coli* [18].

Methylotrophs are diverse organisms that can use reduced one-carbon (C₁) compounds (e.g., methane, methanol) or multi-carbon compounds lacking carbon-carbon bonds (e.g., trimethylamine) as sole sources of carbon and energy. Methylotrophs are of practical importance in cycling C₁ compounds like methane in the environment, consuming methylated compounds that affect microbiome-gut interactions [19–21], and converting C₁ substrates to valuable products in industrial settings [22]. Due to their metabolic capacity, methylotrophs face the unique challenge of managing high fluxes of formaldehyde as a central metabolic intermediate.

Even for methylotrophs, where the metabolic pathways for the production and consumption of formaldehyde have been identified, no genes have been identified that allow cells to sense and respond to formaldehyde. The alphaproteobacterium *Methylophilum extorquens* (formerly
Methylobacterium [23] is the most extensively studied facultative methylotroph. During growth on methanol, M. extorquens uses periplasmic methanol dehydrogenases to oxidize methanol to formaldehyde, which is presumed to diffuse through the inner membrane or possibly undergo active transport [24] (Fig 1). In the cytoplasm formaldehyde condenses with the C$_1$ carrier, dephospho-tetrahydromethanopterin (dH$_4$MPT) [25,26], a reaction catalyzed by formaldehyde-activating enzyme (Fae, EC: 4.2.1.147) [27]. Through a series of dH$_4$MPT intermediates, C$_1$ units are oxidized to formate. This branchpoint metabolite is then either further oxidized to CO$_2$ or assimilated into biomass via a tetrahydrofolate-linked pathway and, subsequently, the serine cycle [28]. The dH$_4$MPT pathway plays the dual role of formaldehyde oxidation for C$_1$ growth and detoxification, whereby mutants with a defective dH$_4$MPT pathway exhibit methanol sensitivity due to an inability to detoxify formaldehyde produced intracellularly [25]. Indeed, at the level of gene expression, Fae and the enzymes of the dH$_4$MPT pathway are present at high levels during growth on either C$_1$ or multi-C growth substrates [29–31]. These data have led to the picture that methylotrophs, which can have intracellular fluxes of 100 mM/min [27] and intracellular concentration of ~ 1 mM [32] during growth on methanol, simply have sufficiently high, constitutive level of formaldehyde oxidation capacity to prevent toxic buildup, perhaps obviating formaldehyde sensing and an associated response.

Recent work demonstrating that formaldehyde stress tolerance is responsive to the environment suggests that cells can sense this toxic intermediate. Formaldehyde tolerance was found to be phenotypically heterogeneous across genetically identical individuals of M. extorquens PA1, whereby some cells could not even survive exposure to 1 mM formaldehyde when it was provided in the growth medium [33]. This is consistent with the paradoxical finding that many methylotrophs are unable to directly grow on formaldehyde even though they rapidly generate it during growth on more reduced C$_1$ compounds [25,33]. Although some of the cells in populations were surprisingly sensitive to formaldehyde, other rare cells could grow at normal rates under conditions where the vast majority rapidly die. The distribution of formaldehyde tolerances was found to rapidly shift upwards in response to formaldehyde stress and the distribution would relax downwards in its absence. Critically, differential death and growth could not explain the distribution shifts, arguing that cells have yet undiscovered systems to sense and respond to
formaldehyde and do not rely solely upon consistently high levels of dH₄MPT pathway enzymes for resistance.

Here we have employed experimental evolution to select for growth on formaldehyde and have uncovered multiple loci encoding genes that can impact formaldehyde resistance. Experimental evolution has unique advantages compared to the classical approach of examining mutants defective in a process [34,35], including the ability to invoke the role of essential genes and gain-of-function mutations, both of which we observed herein. The simplest mechanistic hypothesis for increased resistance would be that – like antibiotic resistance mediated by enzymatic modification – evolved resistance would be mediated by an increase in formaldehyde oxidation. Thus, it came as a surprise that none of the loci with beneficial mutations are known to be related to formaldehyde oxidation or any other known methylootrophy gene [36]. Instead, we identified several novel loci, most commonly a gene of unknown function that we name efgA (enhanced formaldehyde growth). We explored the role and function of EfgA through a combination of X-ray crystallography, molecular modeling, mutational analysis, and biochemical characterization, revealing that EfgA is a sensor that directly binds formaldehyde and leads to an arrest in protein translation. EfgA is beneficial to cells when confronted with elevated levels of internally produced formaldehyde and, through phylogenetic analyses and heterologous expression, we show that EfgA function is broadly conserved in methylotrophs. Furthermore, EfgA-mediated formaldehyde protection is transferable to non-methylotrophs. Our findings represent the first characterized formaldehyde stress response in methylotrophs and demonstrate a unique strategy where a single protein senses a toxic metabolite and leads to inhibition of protein translation.
Materials and methods

Bacterial strains, media, and chemicals

*Methylobacterium* (reclassified from *Methylobacterium* [23]) strains used in this study are derived
from *M. extorquens* PA1 [37,38] lacking cellulose synthesis genes to optimize liquid growth
measurements [39]. Thus, the genotype referred to herein as ‘wild-type’ (CM2730) is more
accurately *ΔcelABC*.

*Escherichia coli* strains used in the physiological studies were derivatives of BW23474 [40] while
those used for cloning and protein overexpression were derivatives of *E. coli* NEB 10-beta and
BL21 (DE3) (Stratagene), respectively.

All strains used in this study are described in Tables 1 and S1.

Growth experiments for *M. extorquens* were performed in a modified Hypho medium [41] or
*Methylobacterium* PIPES (MP) medium [39] with 3.5 mM succinate, 15 mM methanol, or 2, 4, 5,
6, 8, 10 mM formaldehyde as a sole carbon source. For growth on solid medium, Bacto Agar (15
g/L, BD Diagnostics) was added and the concentrations of succinate or methanol were increased
to 15 or 125 mM, respectively, or BD Difco nutrient agar was used.

Formaldehyde stock solutions (1 M) were prepared by boiling 0.3 g paraformaldehyde and 10 mL
of 0.05 N NaOH in a sealed tube for 20 m stocks were kept at room temperature and made fresh
weekly (growth experiments) or daily (*in vitro* binding experiments). When present in the media,
compounds were at the following final concentrations: kanamycin (50 µg/mL); tetracycline (12.5
µg/mL), trimethoprim (10 µg/mL), streptomycin (100 µg/mL), sucrose (50 g/L), cumate (30
µg/mL). Stock solutions of glyoxal, acetaldehyde, butyraldehyde, glutaraldehyde, propionaldehyde were prepared in water or ethanol.

Growth experiments for *E. coli* were performed in MOPS medium [42] with 2 mM glucose, 0.7-
1.1 mM formaldehyde, and 0.5 mM L-rhamnose for induction.

Chemicals were purchased from Sigma Aldrich, VWR or ThermoFisher Scientific.
Growth analyses

Starter cultures of *M. extorquens* (2 mL) were grown in biological triplicate by inoculating media with individual colonies. Cultures were grown in Hypho or MP liquid medium with shaking (250 rpm on platform shaker or 70 rpm in a New Brunswick TC-7 culture roller drum) during incubation at 30 °C. Early stationary-phase cultures (24 h for succinate, 36 h for methanol) were then subcultured (1/64) into relevant media for growth measurements. Cell density was determined by monitoring absorbance with a Spectronic 200 (Thermo Scientific) or a SmartSpec Plus (Bio-Rad) at 600 nm. To determine cell viability (CFU/mL), a 100 µL aliquot of culture was used to harvest cells by centrifugation. The supernatant was discarded and the cell pellet was resuspended into MP medium (no carbon). Cell suspensions were then serially diluted (1/10 dilutions, 200 µL total volume) in 96-well polystyrene plates with MP medium (no carbon) and 10 µL aliquots of each dilution were spotted to MP medium plates (15 mM succinate) in technical triplicate. Plates were inverted and incubated at 30 °C until colony formation was apparent (4-6 d), at which point colonies were counted. Technical triplicates were averaged for each biological replicates and biological replicates were averaged.

Starter cultures for growth analyses of *E. coli* were initiated from freezer stocks of WM8637 or WM8653 into 5 mL tubes at 37 °C shaken at 250 rpm. After growth to stationary phase overnight, these were subcultured (1/500) into MOPS medium with 2 mM glucose. After overnight growth, both cultures were diluted to an OD$_{600}$ = 0.02 into MOPS with 2 mM glucose with or without 0.5 mM rhamnose. A volume of 640 µL was pipetted into Costar 3548 48-well plates (Corning) and grown without lids at 37 °C in Synergy H1 plate readers (BioTek) with double-orbital shaking at 425 rpm and a 3 mm orbit and readings taken every 15 minutes (OD$_{600}$ with 100 ms delay and 8 measurements per data point). After 2.33 h to establish exponential growth, the dispenser of the Synergy H1 was used to automatically deliver formaldehyde at 225 µL/sec from a 32 mM stock made in H$_2$O (volume added ranged from 14-22 µL) straight into the wells to final concentrations of 0.7, 0.9, or 1.1 mM. Data were analyzed by first subtracting the average of the blank wells used at the corners of the plate. The data shown are from non-edge wells on the plate; consistent trends with shifted times of recovery were found for the daily duplicates in the edge wells. The entire experiment was repeated on three additional days (each with duplicates) from separate starter cultures.
Genomic context of *efgA* and *efgB* in other organisms

To examine synteny of *efgA* and *efgB* the genomic context of each gene was examined in organisms with closely related homologs using GeneHood (https://gitlab.com/genehood/genehood-cli) and the MiST3 (Microbial Signal Transduction Database) [43]. Figures were generated using the GeneHood software.

Experimental evolution

From individual colonies, three independent cultures of CM2730 were grown in 10 mL liquid Hypho medium in batch culture in 50 mL flasks (sealed) supplemented with 15 mM methanol. Upon reaching stationary phase, a 156 µL aliquot (1/32 inoculum) was transferred to fresh medium containing decreased methanol and increased formaldehyde concentrations (Fig S1). For the first 60 generations the concentration of formaldehyde increased with each transfer. For the next 90 generations the concentration of formaldehyde was kept constant (20 mM). Final populations (150 generations) were plated to a series of solid media containing: i) nutrient agar, 15 mM methanol, 3.5 mM succinate, 15 mM formate, 5 mM formaldehyde, ii) Hypho, 4 mM succinate, iii) Hypho, 100 mM methanol, and iv) Hypho, 30 mM formate. Colonies that arose were streaked for isolation on the same respective medium and were further characterized.

A second round of experimental evolution used 25 individual colonies and only involved four transfers after an initial round of growth at 15 mM methanol: i) 10 mM methanol, 1 mM formaldehyde, ii) 5 mM methanol, 2.5 mM formaldehyde, iii) 5 mM methanol, 5 mM formaldehyde, and iv) 5mM formaldehyde.

Sequence acquisition and phylogenetics

**Maximum likelihood phylogenetic tree of EfgA.** The amino acid sequence of EfgA was used as a query with PSI-BLAST [44] using an 1E^{-10} cutoff in RefSeq [45]. From each of the 5172 matches, the DUF336 sequence was extracted for analysis; sequences > 30% identity were analyzed with CD-HIT [46], allowing us to eliminate all sequences with 90% or greater identity. The average amino acid size was then calculated for the database, and any sequence longer 1σ was removed prior to alignment. The formaldehyde-bound crystal structure of EfgA 6C0Z, was used
as a query for homologous structures, which were then added into the sequence database manually. The 5172 sequences were then aligned with MUSCLE [47] using default parameters. The alignment file was then analyzed with FastTree 2.1 [48] with the LG + CAT [49], WAG + CAT [50], and JTT + CAT [51] models with and without gamma distribution. The LG + CAT model generated the best LogLk (-584712.707) and bad splits (15/5168). The phylogenetic reconstruction was then analyzed and annotated with iTOL [52] with marker positions for known structures added to the tree. The final phylogenetic data is available at TreeBASE (http://purl.org/phylo/treebase/phylows/study/TB2:S27073).

**Maximum likelihood phylogenetic tree of EfgB.** The nucleotide sequence of EfgB was used as a query with BLASTN [44] using a filter to eliminate all sequences with 90% or greater identity; the top 100 matches (all >65% identity) were used to assess phylogeny. The evolutionary history was inferred by using the maximum likelihood method and the general time reversible model [53]. The bootstrap consensus tree was inferred from 500 replicates [54]. The percentage of replicate trees in which the associated taxa clustered together in the bootstrap test (500 replicates) are shown next to the branches [54]. Initial tree(s) for the heuristic search were obtained automatically by applying Neighbor-Join and BioNJ algorithms to a matrix of pairwise distances estimated using the Maximum Composite Likelihood (MCL) approach, and then selecting the topology with superior log likelihood value. This analysis involved 97 nucleotide sequences. There were a total of 1578 positions in the final dataset. Evolutionary analyses were conducted in MEGA X [55]. The final phylogenetic data is available at TreeBASE (http://purl.org/phylo/treebase/phylows/study/TB2:S27073).

**Genetic approaches in M. extorquens and E. coli**

Allelic exchange was used to introduce changes into relevant genetic loci as previously described [56]. Due to challenges with identifying a tetracycline concentration that would reliably differentiate between strains with and without the resistance marker in *M. extorquens* PA1, we constructed a kanamycin resistant version of the allelic exchange vector. A 2 kb region of pCM433 [56] encoding cat and most of *tet* was excised using *Eco*RV. The remaining vector backbone, with the exception of a 0.9 kb region containing *bla* was PCR amplified and joined via Gibson assembly (HiFi DNA Assembly, New England Biolabs) with a 1.3 kb PCR product containing *kan* PCR
amplified from pCM66 [57] to generate pPS04 (Fig S2). The complete sequence of pPS04 is available in GenBank (Submission 2392028, pending) and has been deposited at Addgene (pending).

Deletions of efgA, eliminated 404 bp from the ORF of *Mext_4158* (21-424/435); deletions of efgB, fmt, and fmt-def eliminated the entire coding region(s) of *Mext_0606* (1401 bp), *Mext_1635* (930 bp), and *Mext_1635-Mext_1636* (1456 bp), respectively.

Inducible expression vectors derived from pLC290 were used to express efgA from *M. extorquens* PA1 and *Mfla_1444* from *Methylobacillus flagellatus* KT in *M. extorquens* (pDN147 and pDN162, respectively). Vectors include 30 bp upstream sequence of each gene including respective native ribosomal binding sites.

The WM8655 strain was generated to express efgA from *M. extorquens* PA1 from the rhamnose-inducible *P_{rhaS}* promoter in *E. coli*. The efgA coding sequence plus 30 bp at the 5’ end was amplified using primers with 30 nt overlaps to permit Gibson assembly (HiFi DNA Assembly, New England Biolabs) into pAH120 [40] that had been digested with *Xba*I and *Nde*I, generating pDN380. An empty control strain, WM8637, was first generated by electroporation of pINT-ts into BW23474 [40]. pDN380 was introduced via electroporation into WM8637 to generate WM8655. Constructs were confirmed by analytical PCR and sequencing of the efgA locus.

All vectors were designed using SnapGene software. The Gibson assembly kit from New England Biolabs was used to construct vectors from restriction enzyme-digested, linearized vector backbone and PCR-generated inserts. For *E. coli*, transformations were performed using standard (WM8637 and WM8655) or manufacturer’s (BL21 (DE3)) protocols for chemical transformation. For complementation of *M. extorquens*, triparental conjugations were performed using pRK2073 [58].

**Formaldehyde quantification**

Formaldehyde concentrations in the culture media were measured as previously described [59]. Supernatant from a 100 µL aliquot of culture was isolated by centrifugation (14,000 x g). In technical triplicate, 10 µL of the supernatant or 100 µL of 0.1X supernatant (diluted with MP
medium, no carbon) was combined with 190 or 100 µL Nash reagent B (2 M ammonium acetate, 50 mM glacial acetic acid, 20 mM acetylacetone), respectively, in 96-well polystyrene plates. Reaction plates were incubated (60 °C, 10 min), cooled to room temp (5 min) and absorbance was read at 432 nm on a Wallac 1420 VICTOR Multilabel reader (Perkin Elmer). Formaldehyde standards were prepared daily from 1 M formaldehyde stock solutions and a standard curve was alongside all sample measurements.

**Crystallization and structure of *M. extorquens* EfgA in complex with formaldehyde.**

**Expression and purification of EfgA from *M. extorquens*.** The gene encoding EfgA (including the thrombin cleavable 6XHis tag) was transformed into *E. coli* strain BL21 (DE3) (Stratagene) for overexpression from pDN79. The protein was overexpressed by growing cells in Luria Bertani broth (LB) medium to an A<sub>600</sub> of 0.6 at 37 °C and subsequent induction with 0.5 mM IPTG for 14 h at 16 °C. The cell pellet was collected by centrifugation and resuspended in buffer A (50 mM Tris pH 7.5, 1 M NaCl, 20 mM imidazole, 0.3 mM DTT, 0.2 mM PMSF and Complete protease inhibitor cocktail tablet, EDTA-free (Roche Diagnostics Corp, Indianapolis, IN, US)). Lysis was performed by sonication. The lysate was then centrifuged for 60 m at 24,000 rpm at 4 °C. The supernatant was then applied onto a Hi Trap affinity (5 ml) (Ni<sup>2+</sup>) column (GE Healthcare Life Sciences) The column was washed with 10 column volumes of buffer A, and the protein was then eluted with increasing concentrations of imidazole from 20 to 500 mM. The fractions containing the protein of interest were pooled, dialyzed against 50 mM Tris pH 7.5, 0.3 mM DTT, 0.5 mM EDTA and purified over a Q-XL Sepharose column (GE Healthcare) using a 0.1-1 M NaCl gradient. The fractions containing EfgA were pooled and concentrated in an Amicon centrifugal filter concentrator with a 10 kDa cutoff membrane (Millipore). The concentrated protein was then further purified by size-exclusion chromatography using a Superdex-200 column (GE Healthcare, HiLoad 16/60) equilibrated with buffer (50 mM Tris pH 8.0, 250 mM NaCl). Again, the fractions containing the protein were pooled and concentrated with an Amicon centrifugal filter concentrator with a 10 kDa cutoff membrane. The purity of the protein was analyzed with 15% SDS–PAGE by using ImageJ software and was determined to be greater than 95%.

**Crystallization and structure.** Crystals of *M. extorquens* EfgA were obtained by sparse matrix screening at 15 mg/mL at 4 °C and 10 °C. Preliminary results were followed by optimization of
the successful condition manually using the sitting drop vapor diffusion method. The best quality crystals were grown in 0.2 M potassium fluoride, 2.2 M ammonium sulfate ((NH₄)₂SO₄) at 10 °C. The EfgA-formaldehyde crystals suitable for data collection were grown at 10 °C in 0.2 M KNO₃, 2.2 M (NH₄)₂SO₄, 16.6 mM formaldehyde using the hanging drop vapor diffusion method. Diffraction data sets for *M. extorquens* EfgA were collected at 100 K at 1.65 Å resolution at the Argonne National Laboratory’s Advanced Photon Source (ANL APS) beamline 21-ID-G on a MarMosaic 300 CCD detector. Diffraction data for EfgA in complex with formaldehyde were collected at 100 K at 1.85 Å resolution at Argonne National Laboratory’s Advanced Photon Source (ANL APS) beamline 21-ID-D on a MarMosaic 300 CCD detector. X-ray diffraction data were processed using HKL2000 [60]. The crystal of apo-protein belonged to the space group P22121 with the unit cell parameters a=71.82, b=72, c=105.18, α=β=γ=90°. *M. extorquens* EfgA in complex with formaldehyde crystallized in space group P21221 with the unit cell parameters a=72.06, b=72.06, c=104.56, α=β=γ=90°. The 3-dimensional structures were determined by molecular replacement using *Klebsiella pneumoniae* protein OrfY (Pfam DUF336) domain as the search model in Phaser-MR [61]. The molecular replacement for EfgA-formaldehyde complex was further confirmed by the initial (2Fo- Fc) map generated using Coot [62] that clearly indicated electron density for the formaldehyde that was not included in the original search model. The structure was refined using the Phenix suite [63] and Coot [62]. Ramachandran plots and root-mean-square deviations (rmsd) from ideality for bond angles and lengths were determined using a structure validation program, MolProbity [64]. A summary of data collection and refinement statistics are listed in Table 2.

**In silico predictions of formaldehyde binding site and folding and binding stabilities of EfgA variants**

To predict the location of the formaldehyde binding pocket on the EfgA tetramer, and to estimate folding and binding stabilities of EfgA variants, classical molecular dynamics (MD) simulations were carried out. These simulations used the apo EfgA tetramer X-ray crystal structure (PDB ID: 6BWS) and EfgA monomer (chain A of EfgA tetramer) and were performed using GROMACS v2018 [65]. The AutoDock Vina program [66] was then used to dock formaldehyde to snapshots of the EfgA tetramer for 100 snapshots extracted from the MD simulations to determine the pockets most heavily populated with high scoring poses. EfgA monomer and tetramer snapshots
were analyzed using FoldX software (MD+FoldX approach) to estimate folding and binding stabilities of EfgA variants.

**Molecular dynamics simulation of EfgA monomer and tetramer.** Both EfgA tetramer and EfgA monomer structures were subjected to atomistic MD simulations using the same protocol. MD simulations were performed using AMBER99SB*-ILDN [67] forcefield. The EfgA structure was placed in a cubic box of TIP3P water and the net charge was neutralized by adding Na⁺ and Cl⁻ ions at a concentration of 0.15 M. Protonation states for all ionizable residues were automatically assigned for neutral pH. The system was then minimized using the steepest descent algorithm for 10,000 steps. The subsequent equilibration process was to perform 1 ns simulation with the positions of all heavy atoms in the complex harmonically restrained to allow equilibration of the water molecules around the proteins, followed by another 1 ns simulation with no restraints.

During equilibration, the temperature and the pressure of the system was set to 300 °K and 1 atm respectively using the Berendsen algorithm [68]. Production simulations were then carried out for 100 ns with pressure maintained using Parrinello-Rahman barostat [69] and temperature was controlled using the v-rescale thermostat [70]. Particle mesh Ewald [71] was used to treat electrostatics with a real-space cutoff of 1.2 nm. Van der Waals interactions were cut off at 1.2 nm with the Potential-shift-Verlet method for smoothing interactions. The LINCS algorithm [72] was applied to constrain all bonds to their ideal lengths and timestep of 2 fs was used. During the 100 ns production simulation snapshots were saved every 1 ns giving 100 snapshots of EfgA tetramer and EfgA monomer to be used for docking calculations and FoldX analysis.

**Docking of formaldehyde to EfgA tetramer.** Each of the 100 EfgA tetramer snapshots obtained during MD simulations was used to dock formaldehyde with the AutoDock Vina software [66]. Generally, docking programs allow the ligand to be completely flexible during the conformational search with only a few restricted side chains on the protein assigned as flexible. Use of 100 snapshots from MD simulation allows us to at least partly overcome this limitation.

The 3-D coordinates of formaldehyde were obtained from PubChem (https://pubchem.ncbi.nlm.nih.gov) and assigned Gasteiger partial charges using Autodock Tools (http://mgltools.scripps.edu/). Ligand docking was then carried out by creating a grid box of size 80 Å × 60 Å × 60 Å, centered on the geometric center of the EfgA tetramer, with a grid spacing of
1 Å. All regions of the tetramer protein complex were included in the search for the most favorable interactions of the ligand. The input exhaustiveness parameter for the docking was set to 400. The number of top docking orientations with high docking scores was fixed to 20. This docking protocol was applied to all 100 snapshots and the X-ray crystal structure of the EfgA tetramer, yielded 2020 (101 snapshots × 20 top docking conformations) conformations of formaldehyde bound to the EfgA tetramer. Highly populated docking clusters were then identified using VolMap plugin built in VMD software [73].

Predicting folding and binding stabilities of EfgA variants. A mutation of EfgA can affect the folding of a monomer and/or the formation of the tetramer. In order to determine how amino acid mutations, including the known variants, alter stabilities (ΔΔG values) for EfgA folding and formation of a tetramer we calculated ΔΔG values of folding and binding using our previously successful MD+FoldX approach [74–76]. This involves analyzing MD snapshots with FoldX software [77]. MD snapshots of the EfgA monomer and tetramer were analyzed using the same protocol reported in our previous study [76]. Briefly, each snapshot was subjected to the RepairPDB command six times in succession to minimize and obtain convergence of the potential energy. For each snapshot, all possible 19 single mutations in the monomer/tetramer at each amino acid site were then generated using BuildModel command. Lastly, the folding stability of the EfgA monomer due to each mutation was estimated using Stability command, and the binding stability of the EfgA tetramer was estimated using AnalyseComplex command. For each mutation, we then estimated ΔΔG of folding and binding by averaging the FoldX results across all individual snapshot estimates. This process led to a total of 2,546 (134 EfgA residues × 19 possible mutations at each site) ΔΔG values for both folding and binding.

Comparison of structural homologs

Structures for EfgA homologs (OrfY – 2A2L, HbpS – 3FPV, Ybr137w – 4CLC, DESPIG_02683 – 4NKP, PduOC – 5CX7, and EfgA – 6BWS and 6C0Z) were acquired from PDB (https://www.rcsb.org/) and aligned with PyMol v2.3 [78].

EfgA:ligand binding assays
Expression and purification of EfgA. Recombinant EfgA containing a C-terminal 6X-His tag was expressed in *E. coli* BL21(DE3) housing pET28 vectors (Novagen). Single colonies from LB supplemented with 50 µg/mL kanamycin were used to generate 25 mL overnight cultures grown at 37 °C with continuous shaking at 250 RPM. Cultures were diluted to an OD$_{600}$ = 0.05 in a 1 L flask containing 800 mL of LB kanamycin then placed at 37 °C. When cultures reached OD$_{600}$ ~ 0.52-0.60, they were induced with 1 mM isopropyl β-D-1-thiogalactopyranoside (IPTG) and grown for 4 hours at 37 °C. Cells were then harvested by centrifugation at 10,000 x g for 20 m at 4 °C. Harvested cells were washed with Buffer A (50 mM Na$_2$HPO$_4$, 300 mM NaCl, 35 mM imidazole at pH 8.0), then harvested again and resuspended in 15 mL Buffer A. Cells were lysed via French Pressed (Thermo Fisher Scientific) in a pre-chilled cell at 20,000 psi. Cell lysate was cleared by centrifugation at 50,000 x g for 2 hours at 4 °C followed by filtration through 0.22 µm PTFE filters and stored at −80 °C prior to purification.

Protein lysates were thawed at 4 °C and centrifuged briefly to ensure no precipitants were present prior to column chromatography. An NGC FPLC (Bio-Rad) was used to purify the C-terminal 6X-His EfgA with 1 mL Ni-NTA columns (Bio-Rad). Columns were equilibrated with 10 mL of Buffer A at 1 mL/min prior to loading lysates. Lysates were loaded at 0.35 mL/min followed by 10 mL of Buffer A at 1 mL/m. An isocratic phase was generated by passing 10 mL of Buffer B (50 mM Na$_2$HPO$_4$, 300 mM NaCl, 500 mM imidazole at pH 8.0) gradually through the column in reverse phase at 0.5 mL/min with samples being collected every 0.5 mL. Degassed H$_2$O (10 mL) was run at 1.0 mL/min until conductivity and absorbance (280 nm) were zero. Samples with high 280 nm values were collected during the isocratic phase of Buffer B.

Fractions from FPLC purification were treated with Laemmli buffer [79] with 10 mM 1,4-dithiothreitol (DTT) and heated for 5 m at 85 °C then analyzed on a 4-15% discontinuous sodium dodecyl sulfate polyacrylamide gel electrophoresis (SDS-PAGE) gel with a 6% stacking gel run at ambient temperature at a constant 100 V. Samples that showed high purity were pooled and stored with 9% glycerol (v/v) and then quantitated with a Bradford assay [80] prior to storage at −80 °C.

Microscale isothermal calorimetry (mITC). Binding of formaldehyde and EfgA was measured via mITC using an Affinity ITC - LV (TA Instruments - Waters LLC). An isothermal buffer for
experimental titrations of EfgA was designed to minimize binding between ligands and buffer (50 mM Na$_2$HPO$_4$, 300 mM NaCl, pH 8.0). Pooled samples of EfgA or EfgA variants were buffer exchanged 1:128 fold with mITC buffer using 3K MWCO Amicon cellulose acetate filters at 4000 x g for 30 m in a swinging bucket rotor. Protein samples were quantitated via Bradford assay and normalized to 2 mg/mL using mITC buffer. Initially, protein:ligand concentrations were used in varying ratios to determine working assay parameters. Proteins were then diluted in mITC buffer to a final concentration of 50 µM. Formaldehyde (1 M) was prepared from paraformaldehyde in milliQ-H$_2$O and used within 24 hours. Formaldehyde stock was serially diluted to 25 mM in mITC buffer. Methanol, formate, and acetaldehyde were all prepared in the same way as formaldehyde to minimize difference between ligand preparation. Prior to use, protein samples and buffer were degassed for 10 m at 650 mm Hg; ligands were degassed for only 5 m to minimize vapor loss.

All experimental runs were performed with 400 µL mITC buffer in the reference and 400 µL sample cells. Between runs, the sample cell and titration syringe were washed 10 times with degassed H$_2$O and 10 times with degassed mITC buffer. The run protocol was 20 injections of 2 µL of 25 mM ligand (with the first being a 0.3 µL throw-away titration) every 200 s with a stir speed of 125 RPM, and 25 °C. Prior to any run or data collection, a slope (µW/h) difference of 0.30 and standard deviation (µW) of 0.03 was required between reference and sample cell.

Isotherms of buffer:ligand were subtracted from protein:ligand data prior to calculation of binding energies. The blank µcal energy was subtracted from the total, which acted as the heat of dilution of ligand into protein.

**Microscale thermophoresis (MST).** MST experiments were performed on a Monolith NT.115 system (Nanotemper Technologies, San Francisco, CA, USA). A solution of unlabeled formaldehyde was serially diluted in reaction buffer (10 mM Na$_2$HPO$_4$, 1.8 mM KH$_2$PO$_4$, 2.7 mM KCl, 137 mM NaCl, 0.05% (v/v) Tween-20) to which an equal volume of Alexa-647 labeled EfgA was added to a final concentration of 20 nM. The samples were loaded into standard treated capillaries (Nanotemper) using 70% LED and 80% IR-laser power. Laser on and off times were set at 30 s and 5 s, respectively. The resulting K$_d$ values are based on an average from three independent MST measurements. Temperature of MST experiments were 20 °C and 30 °C. Data analysis was performed using Nanotemper Analysis software, v.1.5.41.
The raw MST traces for each individual experiment were transformed and fit according to published methods by fitting the normalized fluorescence ($F_{\text{norm}}$) to the Hill equation:

$$F_{\text{norm}} = (F_{\text{norm, max}} - F_{\text{norm, min}}) x [c_A]^n / ([c_A]^n + K_d^n),$$

where $F_{\text{norm, max}}$ = maximal normalized fluorescence, $F_{\text{norm, min}}$ = minimal normalized fluorescence, $[c_A]$ = concentration of protein, $K_d$ = dissociation constant and $n$ = hill coefficient.

**Formaldehyde tolerance distributions**

To compare the distribution of formaldehyde tolerance phenotypes among individual cells in populations of WT and $\Delta efgA$ mutants, formaldehyde tolerance assay plates were prepared as follows: MP medium was prepared with agar, autoclaved, and cooled to 50 °C; then methanol (final concentration: 125 mM) and formaldehyde (to the desired final concentration) were rapidly mixed in, and the agar was poured into 100 mm petri dishes. The dish lids were immediately replaced and plates were cooled on the benchtop. Plates were stored at 4 °C and used within 1 week of pouring. CFU were plated and enumerated as described above.

Cell cultures were grown in MP-methanol medium until stationary phase. They were then subjected to serial 1:10 dilutions in MP to a final dilution of $10^{-6}$. From each of the seven dilutions, three replicates of 10 µL were pipetted onto each MP-methanol-formaldehyde plate to form spots (total: 21 spots per sample per plate type). The spots were allowed to dry briefly in a laminar flow hood, then lids were replaced and plates were stored in plastic bags and incubated at 30 °C for 4 days before colonies were counted. For each replicate set of seven spots, the two highest-dilution spots with countable colonies were enumerated and summed, then multiplied by 1.1 times the lower of the two dilution factors to calculate the original number of colony-forming units (CFU) in the sample. The frequency of tolerant cells at each formaldehyde concentration was then calculated by expressing the number of CFU at that concentration as a proportion of the CFU measured at 0 mM. For each sample, the mean and standard deviation of the three replicate spot series was calculated. To compare the shape of the curves, we measured the rate of the decline of tolerant cells relative to formaldehyde concentration ($\log_{10}$ cells / mM formaldehyde) by fitting an line to the points where frequencies are consistently statistically different (for WT, the last 4 non-zero values; for $\Delta efgA$, the last 4 values).
This method has a limit of detection of $1.65 \times 10^{-7}$ (an abundance of 34 CFU/mL is necessary to observe 1 cell per 30 µL plated, and the total cell population tested was $2 \times 10^8$ CFU/mL; therefore the least-abundant subpopulation that could be detected, disregarding the effects of Poisson distributions at lower $\lambda$, is one with an average frequency of $1.65 \times 10^{-7}$ within the total population).

Although this assay measures the growth of bacterial colonies and not directly that of individual cells, it has been demonstrated to correlate well with single-cell methods of measuring formaldehyde tolerance distribution [33].

**Competition assays**

In biological triplicate, strains were acclimated to growth in Hypho supplemented with 15 mM methanol. In each competition, a test strain was mixed 1:1 (by volume) with a fluorescent reference strain expressing mCherry (CM3841, $efgA^{evo1}$ $efgB^{evo1}$ $\Delta$hpt::$P_{tacA}$-mCherry). This mixture was then used to subculture (1/64 inoculum) into 5 mL of identical fresh medium and grown as described above. The frequencies of fluorescent and non-fluorescent cells were quantified at the start ($F_0$, $t = 0$) and end ($F_1$, $t = h$ when cells reached stationary phase) of the competition experiment using an LSRII flow cytometer (BD, Franklin Lakes, NJ). mCherry was excited at 561 nm and measured at 620/40 nm. For a given sample, at least 1000 cells were counted. An identical set of competitions were set up with 5 mM formaldehyde in place of methanol as the sole carbon source in the growth medium for acclimation and subsequent competition assays. Malthusian fitness values ($W$) relative to the reference strain were calculated by a previously described equation assuming an average of 64-fold size expansion of mixed populations during competitive growth: $W = \log(F_1x64/F_0)/\log((1-F_1)x64/(1-F_0))$; for a 32-fold size expansion in formaldehyde: $W = \log(F_1x32/F_0)/\log((1-F_1)x32/(1-F_0))$ [41].

**In vivo translation assays**

Succinate-growth stationary phase cultures of wild-type (CM2730) and the $\Delta$efgA mutant (CM3745) were inoculated into MP medium with 15 mM succinate in biological triplicate and grown at 30 °C with shaking. At early exponential phase ($OD_{600} = 0.25$), 1 mM $[^{13}C_3]$-methionine (Met) was added to each culture and mixed to homogeneity. Each culture was divided into three aliquots and treated with i) nothing, ii) 5 mM formaldehyde, or iii) 50 µg/mL kanamycin and
immediately returned to incubator. At 0, 20, 40, 60, 90, 180, and 360 min, the optical density was measured and cells from 1 mL of culture were harvested by centrifugation (top speed, 2 m.) for [13CD3]-Met quantification by modification of a previously described method [81]. For time = 0 m., cells were harvested immediately post [%CD3]-Met addition, prior to formaldehyde/kanamycin treatment. Cells were washed with 1 mL of MP medium (no carbon) with 5 s vortex and again harvested by centrifugation. Cell pellets were resuspended in 200 µL of 6 M HCl. The suspension was transferred to a new Eppendorf tube, incubated (105 °C, 18 h), and then dried (lid open, 95 °C, ~20 h). Pellets were resuspended in dimethylformamide (DMF) and the suspension was transferred to a new Eppendorf tube. Amino acids were derivatized with the addition of N-tertbutyldimethylsilyl-N-methyltrifluoroacetamide with 1% (wt/wt) tertbutyldimethyl chlorosilane (TBDMSTFA), incubated (85 °C, 1 h), and then transferred to GC/MS vials for analysis. GC parameters with minor modifications: pressure, 124.5 psi; total flow, 17.9 mL/min; column flow, 1.35 mL/min; column length, 29 m.

GC/MS analysis showed that the peak area ratio of major peaks 218 m/z (standard Met) and 222 m/z ([13CD3]-Met) was well-correlated with the presence of each species in a mixture and was thus used to measure [13CD3]-Met incorporation in cells.
Results

Evolution of *Methylorubrum extorquens* on lethal concentrations of formaldehyde identifies two novel loci, *efgA* and *efgB*, with homologs that are stress response sensors

We evolved *M. extorquens* PA1 [38] for robust growth on formaldehyde in order to identify what loci might be involved in sensing and respond to formaldehyde toxicity. The first of these experiments involved using a steady transition from growth on 15 mM methanol to growth on 20 mM formaldehyde in the first 60 generations; this concentration was maintained to generation 150 (Fig S2). Testing growth of individual isolates confirmed growth at these previously lethal concentrations (Fig S3).

Resequencing the genome of a representative isolate from each of the three populations identified nonsynonymous mutations in two genes of unknown function that were both mutated in each lineage, *Mext_4158* and *Mext_0606*. *Mext_4158* encodes a single domain, 144 amino acid protein with a domain of unknown function (DUF) 336. *Mext_0606* encodes a 466 amino acid protein that encodes a putative adenylate/guanylate cyclase. Neither gene resides in an apparent operon nor are they located near known methylotrophy genes (Fig S4).

Further sequencing in ten additional, randomly picked isolates from each population confirmed the prevalence of nonsynonymous mutations in *Mext_4158* and *Mext_0606* and identified mutations in a few other loci that did not occur in more than one population (Table 1, S1). The repeated occurrence of mutations in *Mext_4158* and *Mext_0606* suggested that these two loci are of particular importance for enhanced formaldehyde growth, thus we renamed them *efgA* and *efgB*, respectively. In a second round of evolution experiments to rapidly obtain growth in 5 mM formaldehyde, isolates from 18 of the 25 populations had mutations in *efgA* (two frameshifts, a 63 bp deletion, the remainder nonsynonymous mutations (Table 1, S1)); no mutations were identified in *efgB*. Subsequent experimentation described below differentiated the respective roles of *efgA* and *efgB* in formaldehyde growth.

Homologs of EfgA are specifically associated with methylotrophic lineages
Phylogenetic analysis demonstrated that DUF336 domains are encoded in all three domains of life and the representatives with structures are broadly dispersed across the tree (Fig 2A). A few DUF336 family members are encoded in gene clusters of well-characterized metabolic pathways [82–86], and some of these DUF336 proteins are localized within bacterial microcompartments, proteinaceous organelles that can confine catabolic processes that involve the generation of toxic, and often volatile aldehydes [83,84,87]. The role of DUF336 domain in these contexts, however, has remained elusive. The only comprehensive studies on bacterial DUF336 function were with HbpS in the Gram-positive bacterium Streptomyces reticuli [88–95]. HbpS localizes to the extracellular face of the cytoplasmic membrane where it senses environmental heme and, in turn, initiates a signaling cascade that mitigates oxidative stress. This raises the possibility that EfgA, despite low levels of sequence similarity to other DUF336 domains (23% identity to HpbS) may also play a role in sensing.

Phylogenetic analyses suggest that EfgA is linked to methylotrophy. Focusing upon close homologs of EfgA, we noted that these were found in a highly supported clade whose members are almost exclusively characterized methylotrophs (Fig 2B). These 51 sequences originate from a broad phylogenetic range including alpha-, beta-, and gammaproteobacteria, and NC10 (e.g., Methylomirabilis oxyfera) clades. However, we noted that some methylotrophic groups, such as those in the genus Bacillus, do not encode an EfgA homolog within this clade.

**Loss of EfgA function necessary and sufficient for growth on formaldehyde**

In order to understand how efgA is involved in formaldehyde growth, we generated a series of mutants and characterized their phenotype in growth media containing varying concentrations of formaldehyde as the primary carbon and energy source. Introduction of efgA<sup>evo</sup> alleles from evolved isolates into the wild-type background was sufficient to enable robust growth in medium containing up to 5 mM formaldehyde (Fig 3, Fig 4). Conversely, restoration of the efgA<sup>WT</sup> allele in an evolved isolate abolished formaldehyde growth, demonstrating the necessity of efgA<sup>evo</sup> for formaldehyde growth (Fig 3). Like the evolved alleles, a ΔefgA in-frame deletion allele in the wild-type background indicated that the EfgA function must be eliminated to permit growth on formaldehyde. These data, as well as the spectrum of mutations obtained for the efgA<sup>evo</sup> alleles indicate that these are loss-of-function mutations.
Further phenotypic analysis of the $\Delta efgA$ in the wild-type background showed that it enabled growth in medium containing 4-8 mM formaldehyde at initial growth rates comparable to those seen on methanol (Fig 4). The $\Delta efgA$ strain also exhibited increased formaldehyde resistance in the presence of an alternative growth substrate (Fig S5). Correspondingly, introduction of a second copy of $efgA^{WT}$ in the chromosome of wild-type resulted in increased sensitivity to formaldehyde in the presence of an alternative growth substrate (Fig S6). These data show that formaldehyde growth was due to increased formaldehyde resistance and $efgA$ plays a key role in the cell’s response to formaldehyde.

**EfgB plays a secondary role in formaldehyde resistance**

$efgB$ encodes a putative adenylate/guanylate cyclase, a group of regulatory proteins that synthesize cyclic nucleotide second messengers. Adenylate/guanylate cyclases are well-known to exist in essentially all organisms and have been associated with a wide variety of phenotypes, ranging from catabolite repression, induction of virulence, and stress response [96–98]. EfgB displays low identity to CyaA of *E. coli* (6.5%) with its most closely related homologs found in other members of the Rhizobiales with diverse physiologies (Fig S7), rather than methylotrophic organisms from a wide variety of phylogenetic groups.

Genetic analyses with $efgB$ alleles indicate that EfgB also plays a role in growth on formaldehyde. Consistent with $efgA^{evo}$ alleles being necessary and sufficient for formaldehyde growth, the introduction of $efgB^{evo}$ alone into the wild-type background did not confer growth on 5 mM formaldehyde (Fig 3). When $efgB^{WT}$ was introduced into an evolved strain to replace the $efgB^{evo1}$ allele, no detectable change in growth was observed in medium containing up to 5 mM formaldehyde (Fig 3), however, when $\Delta efgB$ was introduced into any of the evolved strains growth at higher concentrations of formaldehyde was decreased (Fig S8). These data indicate that the $efgB^{evo}$ alleles are gain-of-function mutations and led us to hypothesize that they could increase formaldehyde resistance in a genomic background that had already achieved formaldehyde growth.

To more clearly elucidate the role of EfgB, we further examined different $efgB$ alleles in the $\Delta efgA$ and wild-type backgrounds. In the $\Delta efgA$ background, growth analysis showed that, at higher
concentrations of formaldehyde (6-10 mM) the state of the efgB allele affected growth in the order efgB<sub>evo</sub><sup>+</sup><sup>evo</sup>efgB<sub>WT</sub><sup>Δ</sup>efgB (Fig 5). In the wild-type background, strains with efgB<sub>evo</sub> alleles exhibited modest resistance to formaldehyde, but only when tested at the low concentration of 2 mM formaldehyde in the presence of a primary growth substrate (Fig S9). These data indicate that the efgB alleles did not impact growth in the absence of formaldehyde stress but do increase formaldehyde resistance.

**EfgA has a formaldehyde-specific role in methylotrophy**

Phenotypic analyses of the ΔefgA mutant indicate the impact of EfgA is specific to formaldehyde stress. During growth on methanol or succinate alone, the ΔefgA mutant phenotype was indistinguishable from wild-type with regard to lag time, growth rate, and final yield (Fig S5). Inhibitory levels of several other aldehydes (glyoxal, acetaldehyde, glutaraldehyde, butyraldehyde, propionaldehyde) did not exhibit a differential effect on ΔefgA strains compared to wild-type (Fig S10). These data suggest that EfgA has a formaldehyde-specific role in the cell.

Analysis of efgB mutants indicate that EfgB is involved in a broad stress response and not limited to formaldehyde stress. None of the efgB alleles tested affected growth on methanol or succinate alone (Fig S9). In contrast to the results with the ΔefgA strain, efgB<sub>evo</sub> alleles conferred modest resistance to a number of aldehydes in addition to formaldehyde (Fig S11A). This broader stress resistance extended beyond aldehydes, as efgB<sub>evo</sub> alleles provided resistance to heat shock as well as a few antibiotics (Fig S11B, C).

Taken together, our data indicate that EfgA has a formaldehyde-specific role in the cell whereas EfgB is involved in resistance to multiple stressors. The specific association of homologs of EfgA, but not EfgB, with methylotrophy further corroborates this formaldehyde-specific role. Our data definitively show that EfgA plays a primary role in formaldehyde resistance unlike EfgB, which plays a secondary role. Correspondingly, we focused our efforts to uncover the biochemical function and role of EfgA.

The crystal structure of EfgA and molecular dynamics simulations suggest a formaldehyde-binding pocket
To further develop hypotheses regarding the biochemical function of EfgA we determined the structure of an N-terminal His-tagged derivative. EfgA diffracted to 1.65 Å resolution (Protein Data Bank (PDB): 6BWS). The tertiary structure of individual protomers is comprised of an antiparallel β-sheet flanked by four antiparallel α-helices in a mixed topology (Fig S12). The packing of the monomers is consistent with a homotetrameric quaternary structure (Fig 6A). No cofactors or metals copurified with EfgA. With these structural data in hand, we employed a MD+FoldX approach [76] to rationalize the efgA⁺⁺ alleles that emerged. This analysis suggested that 11 of the 14 non-synonymous efgA mutations either decrease the stability of monomers or of the interactions between them (all but H2Q, M107I, S114N; Fig S13).

Search for a structural homolog via DALI revealed that the Cα positions of the EfgA monomer have a root-mean-square deviation of 1.95 Å (with 127 atoms aligned) for the Cα positions of the monomeric chain of HbpS from S. reticuli (PDB: 3FPV, Fig 2A). HbpS and EfgA only share 36% amino acid identity and EfgA lacks the twin-arginine translocation signal sequence present in HbpS (Fig S14). We did not expect HbpS and EfgA to have identical functions but were intrigued that EfgA may also function as a stress sensor, specifically hypothesizing that it may sense formaldehyde.

X-ray crystallography of formaldehyde-soaked crystals indicated a specific site where formaldehyde may bind. The structure of the protein with formaldehyde as a potential complexed ligand resolved at 1.83 Å resolution (PDB: 6C0Z). The overall structure was largely unchanged from the apo-protein, but the difference maps comparing the apo and formaldehyde complexes, indicated new electron density in the formaldehyde-soaked crystals. The additional electron density was localized to a specific binding pocket of each monomer. In three EfgA monomers, new densities were modeled as oxydimethanol and in the fourth monomer, the electron density was modeled as formate (Fig 6A). Both oxydimethanol and formate are derivatives of formaldehyde and are structurally homologous. Potential hydrogen bonds were observed between ligands and amino acids that corresponded to S114, D121, and K57 in the native EfgA protein. (Fig 6A inset).

Consistent with the structural data from formaldehyde-soaked crystals, molecular dynamics simulations identified the identical pocket as the likely site of formaldehyde binding. Unbiased
Docking calculations were performed to dock formaldehyde to snapshots obtained from 100 ns molecular dynamics (MD) simulation initiated from the X-ray crystal structure of apo EfgA tetramer. In total, 2020 docked poses were captured and regions of high-density poses were found to correspond with the binding pockets identified during formaldehyde crystal soaks (Fig 6B). In addition to the primary interaction (K57, D121, and S114), these calculations suggest a feasible interaction at R42. The potential importance of the K57-D121-S114 pocket is emphasized by the fact that one of the loss-of-function efgA<sup>evo</sup> alleles (EfgA<sup>S114N</sup>) modifies one of these three proposed binding interactions.

**In vitro EfgA:ligand interaction demonstrates direct, specific binding to formaldehyde**

To directly test the hypothesis that EfgA senses formaldehyde by direct binding, as suggested by the structural and biophysical modeling evidence above, we used two independent biophysical approaches. Microscale isothermal titration calorimetry (mITC) binding isotherms showed that formaldehyde binding for native EfgA was exothermic (ΔH = −22.65 ± 1.16 kcal/mol), suggesting formaldehyde binding was favorable. Boiled EfgA had a 23-fold decrease in the enthalpy of binding and broadened isotherms, suggesting that the strong interaction of formaldehyde with EfgA required the native 3-dimensional structure, rather than non-specific interactions (Fig 7A, S15A). Microscale thermophoresis (MST) was used to both independently confirm the interaction of EfgA with formaldehyde and permit calculation of its affinity (K<sub>d</sub>=8.01±3.5 mM) (Fig S16).

We tested EfgA interactions with two additional categories of alternative ligands that are structurally similar: other C<sub>1</sub> intermediates (methanol and formate) and a longer aldehyde (acetaldehyde). mITC results indicate no evidence of binding to methanol, formate, or acetaldehyde (Fig 1, Fig 7B, Fig S15D-F). Together, these biochemical data confirm the hypothesis that EfgA specifically binds formaldehyde.

**EfgA homologs from methylotrophs are functionally redundant**

Having demonstrated EfgA from *M. extorquens* binds formaldehyde we hypothesized that EfgA homologs found in methylotrophs would have a conserved formaldehyde-sensing function. We attempted to complement a ΔefgA strain of *M. extorquens* (an alphaproteobacterium) with *Mfla_1444*, the corresponding gene from *Methylobacillus flagellatus* KT (a betaproteobacterium).
Mfla_1444 is an EfgA homolog with 67% identity and 76% similarity and represents a member of the other major clade of EfgA sequences in comparison to the EfgA from *M. extorquens* (Fig 2B). Even the basal, uninduced expression of Mfla_1444 complemented a ΔefgA strain and inhibited formaldehyde growth as well as the native gene (Fig 8). These data suggest that the clade of EfgA homologs found throughout methylotrophs sense formaldehyde and are competent in transmitting that signal to conserved downstream components.

**EfgA protects cells from endogenous formaldehyde stress**

The direct interaction of EfgA with formaldehyde and its conserved function across methylotrophs led us to question what selective benefit EfgA may provide to cells. The experimental evolution conditions we used had selected for the removal of EfgA to allow cells to grow on high concentrations of formaldehyde as a sole carbon source. These conditions are unlikely to be ecologically relevant as they were both extreme and involved exogenous formaldehyde rather than endogenous production as an intermediate. Thus, we hypothesized that the advantage of EfgA regulation may be when methylotrophs experience misbalanced intracellular formaldehyde.

Taking advantage of mutants with partial (Δfae) or complete (ΔmptG) lesions in the dH₄MPT pathway to generate internal formaldehyde stress, we determined that EfgA is beneficial under these conditions. ΔefgA Δfae and ΔefgA ΔmptG strains displayed an exacerbated growth defect compared to the corresponding single mutants with *efgA*<sup>WT</sup> when 1 mM methanol was added during growth on succinate (Fig 9). These data showed EfgA is beneficial when cells experience endogenous formaldehyde stress. Furthermore, they indicate that the EfgA-mediated response is independent of the dH₄MPT pathway enzymes and metabolic intermediates.

**EfgA is not linked to phenotypic formaldehyde tolerance**

Given that mutations in *efgA* impact genetic resistance to formaldehyde, we questioned whether it was linked to phenotypic formaldehyde tolerance observed between genetically identical cells [33]. *efgA* was not represented amongst the differentially-expressed genes seen in the environmentally-responsive phenotypic variation in formaldehyde tolerance (Table S2). Additionally, neither *efgB* nor any of the other loci identified via mutations during selection for formaldehyde growth had significant changes, other than a single modest change in a secondary
locus yet to be investigated (Mext_2112, encoding an XRE family transcriptional regulator/shikimate kinase; 1.19-fold change, p-adj=9.28 E-05) in (Table S2). A comparison of the tolerance distributions of wild-type and the ΔefgA mutant showed that a distribution of formaldehyde tolerance was maintained within populations of the ΔefgA mutant but was shifted toward higher tolerance (Fig 10). Notably, the qualitative shape of the distribution altered with the rate of decline being decreased in the ΔefgA mutant (slope: WT=−4.112 log10 cells / mM formaldehyde, ΔefgA=−1.482, p-value=0.0004). Further work will be required to determine whether their protein levels or activities may play any role in this response.

**EfgA can provide protection from formaldehyde in the non-methylotroph *E. coli***

Does EfgA only function in methylotrophs, or might it perhaps provide protection to a heterologous host that is not a methylotroph? Several pieces of evidence led us to hypothesize that this might be possible. First, EfgA senses formaldehyde directly and does not require a functional C1 pathway, so these functions should be possible in a different genomic context. Second, we demonstrated that a phylogenetically distinct EfgA homolog can complement a ΔefgA mutant, suggesting that any downstream signaling system might be broadly conserved. Third, unlike other DUF336 homologs (e.g., *hbpS, pduO, glcC*), there does not appear to be a conserved genomic context for *efgA* across methylotrophs, suggesting it may have been introduced on its own in the history of those lineages.

Heterologous expression of EfgA in *E. coli* (a gammaproteobacterium) demonstrated EfgA can provide increased formaldehyde resistance in a novel organism. *E. coli* grown in minimal MOPS medium with glucose displays sensitivity to >0.7 mM formaldehyde, but strains expressing EfgA showed a decreased lag time across the formaldehyde concentrations tested (Fig 11). The ability of EfgA to mitigate formaldehyde stress in a new organism, in the absence of native methylotrophic machinery, supported our hypothesis. These data also suggest one of two scenarios exist: EfgA acts independently or EfgA elicits a cellular response via interactions with proteins that are present and sufficiently conserved across non-methylotrophs to enable fortuitous interactions to occur.
Alternative loci that permit growth on formaldehyde in the presence of active EfgA include two ribosome-associated proteins.

Despite having a functional EfgA, four other loci were targets beneficial, first step mutations that permitted growth on formaldehyde. To glean information about possible downstream effects of EfgA we resequenced the genomes of isolates from the 7 (of 25) populations that rapidly evolved growth on 5 mM formaldehyde but had an $efgA^{WT}$ allele. This identified four loci: $potG$ ($Mext_4194$, encoding a putative putrescine transporter), $Mext_0925$ (encoding a MarR family transcriptional regulator), $prmA$ ($Mext_4479$, encoding ribosomal protein L11 methyltransferase), and $def$ ($Mext_1636$, encoding peptide deformylase, PDF) (Table 1, S1). An evolved isolate representing each of these four genes was found to have fairly similar fitness (within 10%) to a reconstructed $efgA^{evo}$ mutant strain during growth on 5 mM formaldehyde (Fig S17). While the role of PotG and the MarR-like regulator remain unclear, the finding that mutations in two different genes encoding ribosome-associated proteins (PrmA and PDF) could protect cells from formaldehyde led us to further explore the unexpected connection between formaldehyde stress and translational events.

$N$-formylmethionine pathway contributes to formaldehyde resistance

The observation that variants of PDF allowed formaldehyde growth led us to hypothesize that the $N$-formylmethionine (fMet) pathway plays a role in formaldehyde resistance. During translation PDF removes the formyl moiety from the fMet at the N-terminus of a majority of the nascent peptides in bacteria. We were unable to delete $def$ in wild-type $M$. extorquens, consistent with the finding that $def$ is often individually essential [99–101]. In other bacteria mutants lacking methionyl-tRNA formyltransferase (encoded by fmt) do not synthesize fMet and no longer require PDF [100–103]. In a wild-type background, deletions of $fmt$ alone or the entire $fmt-def$ operon significantly increased sensitivity to formaldehyde (Fig 12). To test whether the lack of fMet precludes the effects of EfgA activity the $\Delta fmt$ and $\Delta fmt-def$ alleles were combined with the $\Delta efgA$ allele. The resulting strains showed that the deletion of $efgA$ significantly increased resistance even in the absence of the fMet pathway (Fig 12). These results indicate that EfgA must act, at least in part, independent of the N-terminal protein formylation pathway, but further implicate protein translation/maturation as having a key role in formaldehyde resistance.
EfgA halts translation *in vivo* in response to formaldehyde

Multiple findings led us to directly test whether EfgA acts by modulating translation. Translation was assayed *in vivo* by tracking the incorporation of exogenously-provided $[^{13}\text{C}_3]$-methionine (Met) into cellular proteins of wild-type and an ΔefgA mutant by GC/MS. Treating cells with kanamycin, a known translation inhibitor, reduced methionine incorporation by the 360 m timepoint in both genotypes (Fig 13). In contrast, formaldehyde-treated wild-type samples halted translation immediately, showing no detectable increase in methionine incorporation with statistical significance for the full 360 m (Fig 13). For the formaldehyde-treated ΔefgA mutant, translation was not halted and increased between all timepoints. The effects of treatments upon translation were largely mirrored in growth; kanamycin inhibition being slow and formaldehyde inhibition in the presence of EfgA being rapid (Fig S18). The addition of formaldehyde to the ΔefgA mutant did not induce growth arrest and only led to a modest growth defect. These data indicate that, although formaldehyde does exert some inhibition of translation and growth on its own, the primary effect of excess formaldehyde upon translation and growth is mediated by EfgA.
Discussion

Formaldehyde detoxification systems have been identified in all domains of life [104].
Methylotrophs generate formaldehyde at high rate, yet several decades of work with methylotrophs had failed to reveal any proteins that sense and respond to its toxicity. By evolving *M. extorquens* to grow on 20 mM formaldehyde as a sole substrate, a concentration that was found to sterilize a wild-type culture in just two hours [33], we uncovered several genes that could permit growth. Notably, we report here that one of these, now named *efgA* for enhanced formaldehyde growth, encodes a formaldehyde sensor that can act through inhibition of protein translation.

We propose that the role of EfgA is to impart protective growth stasis when formaldehyde is transiently elevated to protect from formaldehyde damage. Metabolite-mediated translational arrest has been previously reported for metabolites involved in translation itself, such as amino acids or ppGpp, but not for endogenous stressors [105]. Such a mechanism could limit formaldehyde-induced damage to cellular components, spare protein resources under stress conditions [106], or even reduce further enzymatic production of formaldehyde during high formaldehyde exposure. Nascent peptides emerging from ribosomes may be amongst the most consequential sites of damage and toxicity of formaldehyde crosslinking. Although the mechanism of EfgA inhibition of translation is unclear, the consequent inhibition may be effective because it provides a rapid mechanism for inducing growth arrest. Sustained high external concentrations of formaldehyde that induced permanent growth arrest in wild-type are quite unlikely to be encountered in nature, however, the ability of EfgA to help cells navigate elevated internal formaldehyde stress may be critical in a fluctuating environment. *M. extorquens* are commonly isolated as key members of the leaf microbiome. Although formaldehyde concentrations in plant tissues are relatively low (0.1-10 µmol/g in *Zea mays* and 4.2 µmol/g in *Arabidopsis thaliana*) [107,108], methanol is generated and released in pulses each day [109,110]. This raises the possibility that the beneficial role of EfgA is to mitigate formaldehyde toxicity when formaldehyde production from MDH and usage by the dH₄MPT pathway become transiently imbalanced. Such imbalances might also arise from metabolic perturbations that can be caused by stressors [111,112], metabolic crosstalk [113], transcriptional bursts [114], or nutrient limitation/shifts [29,115].
Both mITC and MST results confirmed that EfgA binds formaldehyde (Fig 7, S15), but with a relatively high $K_d$ of 8 mM (Fig S16). Given that the intracellular concentration of formaldehyde during growth of $M. \textit{extorquens}$ during steady-state growth on methanol has been estimated to be 1 mM [32], a high $K_d$ above that concentration would ensure that EfgA occupancy would remain unsaturated during methylotrophic growth, leaving the opportunity for a dynamic response when formaldehyde concentrations rise further. Furthermore, the specificity of EfgA binding of formaldehyde (Fig 7, S15) would render EfgA insensitive to the free $C_1$ intermediates upstream and downstream of it, as well as acetaldehyde that would be generated during growth on $C_2$ compounds, such as ethanol.

Curiously, the additional electron densities that arose in the formaldehyde-soaked crystals were more consistent with the formaldehyde derivatives, oxydimethanol and formate and not formaldehyde itself. Each of these compounds are derived from formaldehyde and can spontaneously form in aqueous formaldehyde solutions [116]. Aqueous solutions of formaldehyde are composed of formaldehyde and formaldehyde-water derivatives (methylene glycol and oxydimethanol) that exist in equilibrium. Formaldehyde crystal soaks were performed at 10 °C, where the molar ratio of oxydimethanol is nearly at its maximum (0.6 at 8.3°C) [116] and the molar ratio of formaldehyde is ~ 0.3. Therefore, we suspect that the presence of oxydimethanol was an artifact of the experimental conditions but, in fact, its localization is the physiologically relevant site of formaldehyde in the binding pocket. Alternatively, it is possible that EfgA does not sense the unmodified form of formaldehyde directly, but rather an adduct/derivative only formed when formaldehyde is present at sufficiently high concentrations (e.g., oxydimethanol) and which could serve as a proxy for formaldehyde itself. The presence of formate in one of the four monomers may represent an oxidation product formed during crystallization; the lack of interaction seen via mITC when testing formate directly argue against it being the genuine ligand in vivo. Taken together, our results validated the protein:ligand interaction as distinct from non-specific binding that might be expected from the ability of formaldehyde to nonspecifically form adducts on and crosslink amino acids residues.

Both X-ray crystallography and in silico approaches to dock formaldehyde support the hypothesis that formaldehyde binding demonstrated empirically occurs in a pocket formed via hydrogen bonds with S114, D121 and K57. Fortuitously, a substitution in one of these sites, EfgA$^{S114N}$, was
isolated as one of the loss-of-function alleles isolated from one of the evolving populations. Combined with the high degree of conservation at these three residues (Fig S14), these are top candidates for residues involved in ligand binding and will be investigated further in future work.

The discovery that EfgA is a formaldehyde sensor indicates a previously unknown role for DUF336 proteins in sensing small metabolites. Proteins with DUF336 domains are present in single- or multi-domain proteins and their genetic association with gene clusters that encode well-characterized pathways such as glycolate utilization in *E. coli* (GlcG), 1,2-propanediol utilization in *Salmonella enterica* (PduO), and chloroaniline degradation in *Diaphorobacter* sp. PCA039 (OrfU2) has been noted [82–86]. The two best characterized examples are HbpS of *S. reticuli* and PduO of *S. enterica*. HbpS is located extracellularly where it senses and degrades heme and activates a two-component system involved in oxidative stress [88,92,117,118]. PduO is localized to the pdu microcompartment its DUF336 domain is fused to an ATP: cob(I)alamin adenosyltransferase domain; its DUF336 domain is not required for the activity of the ATP: Cob(I)alamin adenosyltransferase domain *in vitro* but is required for optimal 1,2-propanediol utilization for unknown reasons [86]. Both HbpS and PduO have been shown to bind heme and cobalamin but their cellular locations are distinct, consistent with their disparate functions. In this light, EfgA represents the third biochemically-characterized bacterial DUF336 protein and is distinct from HbpS in that it is cytoplasmic, senses formaldehyde, and modulates translation. EfgA and HbpS overlap in terms of being sensors; EfgA and PduO may exhibit functional overlap due to the involvement of an aldehyde in propanediol utilization. Thus, our work, which assigns a novel, aldehyde sensing function for a DUF336 protein, helps to define the broader role of DUF336 domains.

The involvement of several genes with known involvement in translation suggests that controlling protein damage may be particularly critical for cells confronted with formaldehyde stress. Two secondary mutations that we have yet to examine also suggest the key role for protein quality control in formaldehyde resistance. For *hrcA*<sup>evo</sup>, the apparent loss-of-function mutation would eliminate HrcA, a heat-inducible transcriptional repressor that negatively regulates heat shock genes [119,120]. The intergenic *Mext*<sub>3827</sub>/*Mext*<sub>3828</sub> mutation is upstream of *Mext*<sub>3828</sub>, annotated as encoding HdeA, a periplasmic chaperone protein that prevents aggregation of periplasmic proteins [121,122]. These mutations suggest that preventing/repairing protein damage
may be important for formaldehyde resistance and imply that unchecked formaldehyde stress leads to protein damage in *M. extorquens*. Though formaldehyde damage is most commonly associated with DNA damage, there is precedent for formaldehyde-induced protein damage [13,123] which was also suggested by our previous work [33].

While the catalytic activity of PDF has long been known, it has only recently been identified as a key player in protein quality control [7,124,125]. Typically, as the fMet of an elongating peptide chain emerges from the ribosome exit tunnel, PDF quickly removes its formyl group, making it a suitable substrate for the downstream processing enzyme. In instances where the elongating peptide is misfolded, fMet is less accessible to PDF and serves as a signal for protein degradation [124]. We isolated def<sup>vα</sup> alleles that, through an unknown mechanism, increase formaldehyde resistance, whereas Δfmt and Δfmt-def strains with no fMet cycling decreased formaldehyde resistance. To our knowledge, this is the first indication that fMet modification plays a role in formaldehyde resistance.

The surprising ability for EfgA homologs to influence formaldehyde resistance when introduced between distantly related organisms raises an intriguing possibility that EfgA may directly interact with one or more ribosomal or ribosome-associated components to halt translation. There is precedent for a DUF336 protein to physically interact with ribosome-associated proteins. The homolog in *Saccharomyces cerevisiae*, Ybr137wp, has been shown to be involved in post-translational control, whereby it binds Sgt2 and Get proteins and is involved in the co-translational targeting of tail-anchored proteins into membranes [126]. If EfgA interacts with ribosomal or ribosome-associated proteins, it should be noted that molecular composition of ribosomes and the sequences of the molecular components are amongst most well-conserved aspects of bacteria. The finding that EfgA from *M. extorquens* can provide protection from formaldehyde in *E. coli* may rely upon the fact that its interaction partners were already encoded and expressed there. Having downstream partners already present in genomes could also explain how *efgA* could be acquired by horizontal gene transfer and integrated into the genome as isolated genes without conserved synteny. This would be quite distinct from the rampant exchange of methylotrophy pathways by horizontal gene transfer which appear to have been cointroduced as large, genomically-clustered modules [127]. The ability for EfgA to provide an immediate benefit in dealing with formaldehyde stress may also have more immediate biotechnological benefits. Given that formaldehyde toxicity...
was also a key challenge in the engineering and evolution of an *E. coli* strain that can grow on methanol as a sole carbon source [128], this raises the possibility that introducing EfgA would increase the cells’ ability to grow while producing formaldehyde as a high flux intermediate. We perhaps should not be surprised that metabolic pathways that generate toxic intermediates need molecular systems to sense their accumulation and mount responses that either eliminate the toxin, increase the ability to repair such damage, or help the cell avoid making the molecules that the toxin damages in the first place.
Table 1. Evolved beneficial alleles

| PRIMARY MUTATIONS |
|-------------------|
| Allele            | Position | Mutation | Annotation |
| **Mext_4158**     |          |          |            |
| Protein of unknown function DUF336 (EfgA) |          |          |            |
| efgA<sup>evo1</sup> | 4627319  | G→A      | T60I       |
| efgA<sup>evo2</sup> | 4627352  | C→T      | G49D       |
| efgA<sup>evo3</sup> | 4627304  | C→T      | G65D       |
| efgA<sup>evo4</sup> | 4627151  | Δ11 bp   | frameshift |
| efgA<sup>evo5</sup> | 4627109  | Δ63      | deletion   |
| efgA<sup>evo6</sup> | 4627487  | C→T      | T4I        |
| efgA<sup>evo7</sup> | 4627177  | C→T      | M107I      |
| efgA<sup>evo8</sup> | 4627428  | A→C      | T24P       |
| efgA<sup>evo9</sup> | 4627492  | C→G      | H2Q        |
| efgA<sup>evo10</sup> | 4627157 | C→T      | S114N      |
| efgA<sup>evo11</sup> | 4627226  | T→A      | L91H       |
| efgA<sup>evo12</sup> | 4627275  | T→C      | S75P       |
| efgA<sup>evo13</sup> | 4627170  | G→C      | A110P      |
| efgA<sup>evo14</sup> | 4627223  | Δ1 bp    | frameshift |
| efgA<sup>evo15</sup> | 4627223  | Δ1 bp    | frameshift |
| efgA<sup>evo16</sup> | 4627173  | G→T      | G109C      |
| efgA<sup>evo17</sup> | 4627305  | G→A      | G65S       |
| **Mext_1636**     |          |          |            |
| Peptide deformylase (PDF) |          |          |            |
| def<sup>evo1</sup>  | 1824018  | C→T      | G143S      |
| def<sup>evo2</sup>  | 1824284  | A→C      | V54G       |
| **Mext_0925**     |          |          |            |
| MarR family regulator |          |          |            |
| Mext_0925<sup>evo1</sup> | 1001159  | C→G      | H119D      |
| Mext_0925<sup>evo2</sup> | 1001376  | Δ1 bp    | frameshift |
| **Mext_4194**     |          |          |            |
| ABC transporter-like protein |          |          |            |
| potG<sup>evo1</sup> | 4664528  | C→T      | R115Q      |
| **Mext_4478/Mext_4479** |          |          |            |
| Protein/Enzyme | Accession | Mutation | Description |
|---------------|-----------|----------|-------------|
| Hypothetical protein/ribosomal L11 methyltransferase | prmA<sub>evo1</sub> | 4996712 G→C | -541/-77 |
| SECONDARY MUTATIONS | | | |
| Heat-inducible transcription repressor HrcA | hrcA<sub>evo1</sub> | 443917 Δ1bp frameshift | |
| Adenylate cyclase | efgB<sub>evo1</sub> | 666660 C→T | R332C |
| | efgB<sub>evo2</sub> | 666603 G→A | E313K |
| Molybdenum cofactor cytidyltransferase | Mext_1058<sup>evo1</sup> | 1156254 G→C | G91R |
| XRE family transcriptional regulator/shikimate kinase | Mext_2112<sup>evo1</sup> | 2369958 A→G | L105P |
| P-type Cu<sup>+</sup> transporter | Mext_2690<sup>evo1</sup> | 3008713 G→C | G610R |
| Serine protease Do/acid stress chaperone HdeA | mntR<sub>evo1</sub> | 3975117 T→C | T108A |
| Serine protease Do/acid stress chaperone HdeA | Mext_3827/Mext_3828<sup>evo1</sup> | 4252124 T→C | +371/-293 |
| Cytochrome o ubiquinol oxidase/transmembrane transporter protein | cyoA<sub>evo1</sub> | 4822969 Δ37 bp | -54/+105 |
| Beta-lactamase domain protein | Mext_4456<sup>evo1</sup> | 4963821 A→G | Y265H |
Table 2. Data collection and refinement statistics

|                         | EfgA<sup>WT</sup> | EfgA<sup>WT</sup>/soaking with Formaldehyde |
|-------------------------|-------------------|--------------------------------------------|
| **Data set**            |                   |                                            |
|                         |                   |                                            |
| **Data collection**     |                   |                                            |
| Wavelength (Å)          | 1.000             | 0.97934                                    |
| Resolution (Å)<sup>a</sup> | 42.5-1.6 (1.7-1.6) | 42.3-1.8 (1.89-1.83)                      |
| Space group             | P22121            | P21221                                     |
| Cell dimensions         |                   |                                            |
| a,b,c; (Å)              | 71.82, 72.0, 105.18 | 72.06, 72.06, 104.55                      |
| α, β, γ; (°)            | 90, 90, 90        | 90, 90, 90                                |
| Molecules per a.u.      | 4                 | 4                                          |
| Unique reflections<sup>a</sup> | 62821 (6085)     | 47983 (4714)                              |
| Average redundancy<sup>a</sup> | 12.9 (12.5)    | 11.6 (11.4)                               |
| Completeness (%)<sup>a</sup> | 94.7 (95.5)     | 98.37 (97.32)                             |
| R<sub>merge</sub>%<sup>a,b</sup> | 13.6 (92.2)    | 11.9 (54.5)                               |
| Output <I/sigI><sup>a</sup> | 19.8 (1.8)       | 26.1 (4.7)                                |
| **Refinement**          |                   |                                            |
| R<sub>work</sub>%<sup>c</sup> | 22.15 (28.15)  | 21.59 (23.06)                             |
| R<sub>free</sub>%<sup>d</sup> | 24.53 (29.94)  | 25.46 (26.41)                             |
| r.m.s.d.<sup>e</sup> from ideality |             |                                            |
| Bonds (Å)               | 0.003             | 0.009                                      |
| Angles (°)              | 0.571             | 0.94                                       |
| Average B-factor (Å<sup>2</sup>) | 20.82         | 25.93                                      |
| Ramachandran<sup>f</sup> |                   |                                            |
| Favored (%)             | 96.30             | 98.33                                      |
| Allowed (%)             | 3.70              | 1.67                                       |
| Outliers (%)            | 0                 | 0                                          |
| PDB ID                  | 6BWS              | 6COZ                                       |

<sup>a</sup> Values for the last shell are in parenthesis
$$R_{\text{merge}} = \frac{\sum |I - \langle I \rangle|}{\Sigma I}, \text{where I is measured intensity for reflections with indices of hkl}$$

$$R_{\text{work}} = \frac{\sum |F_o - F_c|}{\Sigma |F_o|} \text{ for all data with } F_o > 2 \sigma(F_o) \text{ excluding data to calculate } R_{\text{free}}$$

$$R_{\text{free}} = \frac{\sum |F_o - F_c|}{\Sigma |F_o|} \text{ for all data with } F_o > 2 \sigma(F_o) \text{ excluded from refinement.}$$

Root mean square deviation

Calculated by using MolProbity [57]
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Figure 1. Methanol utilization pathway in *M. extorquens*.

Methanol is oxidized to formaldehyde by methanol dehydrogenase (MDH). Fae then condenses free formaldehyde and dephosphotetrahydromethanopterin (dH₄MPT). MptG is required for dH₄MPT biosynthesis. The pathway branches at formate which can be further oxidized to CO₂ or routed to the assimilation pathways (e.g., serine cycle). Alternatively, free formaldehyde can bind EfgA and lead to cessation of translation. Our working model is that EfgA prevents formaldehyde-induced protein damage.
Figure 2: Phylogenetic analysis of the DUF336 superfamily indicates close homologs to EfgA are present in a broad array of methylotrophs.

(A) The evolutionary relationship of the DUF336 region was compared to all current amino acid sequences and structures currently on NCBI via maximum likelihood. Solved structures of EfgA homologs are represented on the tree with their respective PDB accession codes and overlaid onto EfgA structure (green). The red colored sequences represent members of the EfgA clade.

(B) Expanded view of the EfgA clade with bootstrap values above 50 highlighted by color according to the key. Bolded names indicate experimentally verified EfgA-like function.

Scale bars for both (A) and (B) indicate substitutions per residue.

The phylogenetic data is available at TreeBASE (http://purl.org/phylo/treebase/phylows/study/TB2:S27073).
Figure 3. Evolved alleles of *efgA* are necessary and sufficient to confer growth on formaldehyde.

Final yields (24 h) of strains grown in liquid MP medium with 5 mM exogenous formaldehyde provided as a sole source of carbon and energy are shown. The wild-type (CM2730) was grown alongside isolates evolved to grow on 20 mM formaldehyde: CM3035 (pink), CM3039 (orange), CM3040 (green), CM3044 (violet). Genetic derivatives of CM3044 where the WT allele replaced the evolved (“evo”) allele are shown with violet and white horizontal (*efgB<sup>WT</sup>*. CM3781) or vertical (*efgA<sup>WT</sup>*. CM3791) bars. Genetic derivatives of wild-type were made by introducing distinct *efgA<sup>evo</sup>* alleles (CM3793, CM3795, and CM3797) or *efgB<sup>evo</sup>* alleles (CM3783, CM3837) present in formaldehyde-evolved isolates; colors are consistent with evolved isolate that alleles were isolated from.
Figure 4. Evolved efgA allele or deletion of efgA recapitulates formaldehyde growth.

(A) Evolved isolate (CM3035, pink), reconstructed efgA\textsuperscript{evo1} mutant (CM3793, red), and ΔefgA mutant (CM3745, black triangles) were grown in liquid Hypho medium with 5 mM exogenous formaldehyde as a sole source of carbon and energy. (B) The ΔefgA mutant strain was grown in liquid MP medium with 0, 4, 6, 8, or 10 mM exogenous formaldehyde as a sole source of carbon and energy. For comparison, representative growth with 15 mM methanol (no formaldehyde) is shown (‘M’). Error bars represent the standard error of the mean of three biological replicates.
Figure 5. Formaldehyde resistance of *efgA efgB* double mutants indicates *efgB<sup>evo1</sup>* is a gain-of-function allele.

The Δ*efgA* (blue triangles), Δ*efgA* Δ*efgB* (green triangles), and Δ*efgA* *efgB<sup>evo1</sup>* mutants (cyan diamonds) were grown in liquid MP medium with 15 mM succinate and 4, 6, 8, 10, or 12 mM exogenous formaldehyde. Error bars represent the standard error of the mean for three biological replicates.
Figure 6. X-ray crystal structure of EfgA and predicted formaldehyde binding sites.

(A) EfgA tetramer with each chain highlighted in different colors. Inset: enlarged view of the EfgA binding pocket showing three crucial amino acids, K57, D121, and S114, potentially involved in H-bond interaction with formaldehyde in the crystal structure.

(B) EfgA tetramer with each chain highlighted in shades of grey with all 2020 docked poses of formaldehyde and two key amino acids, K57 and S114. Each pose is represented by a single carbon atom (black points) of formaldehyde. Blue clusters demonstrate highly populated regions. Orange circles show an enlarged view of each distinct site. Residues at each site are elementally color-coded, with carbon=green, oxygen=red, nitrogen=blue, and hydrogen=white.
Figure 7. Microscale isothermal calorimetry indicates EfgA binds formaldehyde but not methanol, formate, or acetaldehyde.
The binding isotherms represented as heat change (µJ/s) upon injection over time are in the top portion of the split graph, with independent binding modelling on the bottom portion. (A) Binding observed with 50 µM EfgA (black) and 2 µL injections of 25 mM formaldehyde (in PBS). (B) Binding observed with 50 µM EfgA and 2 µL injections of 25 mM methanol (blue), formate (green), acetaldehyde (pink). Data are representative of trends observed in multiple experiments (n = 3); additional replicates are shown in Figure S15.
Figure 8. EfgA from *Methyllobacillus flagellatus* complements the *M. extorquens* ΔefgA mutant.

Derivatives of the ΔefgA mutant containing pLC290 were grown in liquid MP medium with 8 mM formaldehyde as a sole source of carbon and energy. Strains contained expression plasmids that were either an empty vector control (CM4625, EV, black), expressed EfgA from *M. extorquens* (CM4126, efgA, blue), or the homolog from *M. flagellatus* (CM4182, Mfla_1444, green). Error bars represent the standard error of the mean for four biological replicates.
Figure 9. EfgA provides protection from internal formaldehyde in methanol sensitive mutants.

Wild-type and mutant strains were grown to early exponential phase in liquid MP medium (succinate) at which point 1 mM methanol was introduced into the medium (t=0 h). Strains represented are wild-type (black circles), Δfae (CM3753, orange squares), ΔmptG (CM4765, light blue squares), ΔefgA Δfae (CM3421-5, green hexagons), and ΔefgA ΔmptG mutants (CM3440-13, blue hexagons). Error bars represent the standard error of the mean for three biological replicates.
Figure 10. EfgA influences phenotypic formaldehyde tolerance distribution in *M. extorquens*.

The distribution of formaldehyde tolerances among individual cells was assessed in WT (black solid circles) and ΔefgA (blue triangles) populations. Stationary-phase cultures were plated onto MP-methanol agar medium containing a range of formaldehyde concentrations at 1 mM intervals. The frequency of tolerant cells is expressed as the ratio of the colony-forming units (CFU) enumerated at the given formaldehyde concentration to the CFU enumerated on formaldehyde-free (0 mM) medium. Error bars represent the mean standard deviation of three replicate plates; the horizontal dotted line denotes the limit of detection.
Figure 11. EfgA provides protection to *E. coli* strains during formaldehyde exposure.

Growth of wild-type *E. coli* and was quantified in liquid MOPS medium with 2 mM glucose and 0, 0.7, 0.9, 1.0, and 1.1 mM exogenous formaldehyde. Strain WM8637 (filled symbols) harboring a chromosomal insert of *efgA* under the control of P*rhaS* or the empty vector control (WM8653, empty symbols) was grown in comparable conditions. *efgA* expression was induced with 0.5 mM rhamnose. Data are representative of trends observed in multiple experiments.
Figure 12. EfgA-mediated formaldehyde resistance is not dependent on PDF. Disc-diffusion assays were performed by placing formaldehyde-impregnated discs upon soft agar overlays of *M. extorquens* on solid MP media (125 mM methanol). The zones of inhibition observed showed that like wild-type, mutants lacking the fMet-mediated protein degradation signal (Δfmt, Δfmt-def) are more formaldehyde-resistant when *efgA* is also deleted. Wild-type and ΔefgA mutant were included as experimental controls. Error bars represent the standard error of the mean for three biological replicates. Statistical significance was determined by an unpaired Student’s *t* test (*, p<0.05).
Figure 13. EfgA mediates rapid cessation of translation in response to formaldehyde.

Exponential phase cultures of wild-type (circles, A) and ΔefgA mutant (triangles, B) strains were treated with kanamycin (purple), exogenous formaldehyde (red) or left untreated (black). After treatment (t >0 min) in vivo translation was assayed via [13CD₃]-methionine incorporation. Error bars represent the standard error of the mean for three biological replicates.