Characterisation of acetogen formatotrophic potential using *E. limosum*

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Key words

Abstract

Formate is a promising energy carrier that could be used to transport renewable electricity. Some acetogenic bacteria, such as *Eubacterium limosum*, have the native ability to utilise formate as a sole substrate for growth, which has sparked interest in the biotechnology industry. However, formatotrophic metabolism in acetogens is poorly understood, and a systems-level characterization in continuous cultures is yet to be reported. Here we present the first steady-state dataset for *E. limosum* formatotrophic growth. At a defined dilution rate of 0.4 d^{-1}, there was a high specific uptake rate of formate (280±56 mmol/gDCW/d), however, most carbon went to CO_2 (150±11 mmol/gDCW/d). Compared to methylotrophic growth, protein differential expression data and intracellular metabolomics revealed several key features of formate metabolism. Upregulation of pta appears to be a futile attempt of cells to produce acetate as the major product. Instead, a cellular energy limitation resulted in the accumulation of intracellular pyruvate and upregulation of Pfl to convert formate to pyruvate. Therefore, metabolism is controlled, at least partially, at the protein expression level, an unusual feature for an acetogen. We anticipate that formate could be an important one-carbon substrate for acetogens to produce chemicals rich in pyruvate, a metabolite generally in low abundance during syngas growth.

Introduction

Acetogens hold great promise for sustainable chemical and fuel production whilst closing the carbon cycle using feedstocks from renewable sources (Ljungdahl 2009). Acetogens can use reducing equivalents through intermediates such as hydrogen (H_2), carbon monoxide (CO), formate and methanol. However, compared to synthesis gas fermentation (syngas, a mixture of CO and H_2), as

has been commercialised by LanzaTech using offgases from the steel industry, much less is known

about the liquid C_1 feedstock, formate (Cotton et al. 2020; Köpke and Simpson 2020).

Unlike the other liquid C_1 feedstock methanol, formate can be efficiently produced directly from CO_2

and renewable electricity without the need for hydrogen (*i.e.* Power-to-X), a technology that is at the

pre-commercialisation stage (Spurgeon and Kumar 2018; Rabiee et al. 2021). Being a liquid, it avoids

many of the transportation issues present with gaseous substrates, as well as fitting with existing

supply chain and fermentation infrastructure (Cotton et al. 2020). It also can overcome key mass

transfer limitations faced by gas fermentation, and have higher energy efficiencies, resulting in lower

operational costs, *e.g.* for like cooling (Cotton et al. 2020). Formate could, in the future, become not

only a biotechnology feedstock, but an energy carrier that avoids the pitfalls of hydrogen such as

non-negligible supply chain leaks leading to additional costs and contribution to global warming as a

result of its GWP over 100 years of 11 (that is, 11 times that of CO_2) (Al-Breiki and Bicer 2020;

Warwick et al. 2022).

Consequently, there have been numerous efforts to engineer formatotrophy into several organisms.

For example, the synthetic reductive glycine pathway has been engineered into *E. coli* (Kim et al.

2020). The so-called FORCE pathway, an orthogonal chain elongation metabolism to biomass based

on formyl-CoA, has also been engineered into *E. coli* (Chou et al. 2021). Despite these advances,

native acetogen formatotrophy has the highest energy efficiency, which may be a critical metric

when considering it as a Power-to-X technology (Neuendorf et al. 2021). Utilising the native

capabilities of acetogens such as *Eubacterium limosum*, *Butyribacterium methylotrophicum* and

Acetobacterium woodii, would avoid the need for genetic engineering or building new gas

fermentation infrastructure.

Formate as a substrate in the Wood-Ljungdahl Pathway (WLP) is similar to CO metabolism in that

there is excess oxidation to CO_2 in order to generate the required number of reducing equivalents.

We have previously shown for *E. limosum* in batch culture that formate consumption results in

acetate and CO_2 production at a stoichiometric ratio of 1:1 (Wood et al. 2021). Interestingly, the

observed native formatotrophic acetogen maximum growth rate was similar to that seen for the

synthetic reduction glycine pathway in *E. coli* (*ca.* 0.1 h^{-1}) (Kim et al. 2020).

Analytical quantification and metabolic modelling have advanced knowledge of acetogen

metabolism, allowing for improved process economics and fermentation designs (Marcellin et al.

2016; Heffernan et al. 2022). Transcriptome and translatome data have been analysed for *E.*

limosum (Song et al. 2017, 2018), and proteomic data for closely related *E. callanderi* (Kim et al.

2021). Some information has been published for the model acetogen, *A. woodii* (Neuendorf et al.

2021), yet no formatotrophic omics datasets exist for *E. limosum* to our knowledge. Since acetogen

metabolism is generally acknowledged to be controlled post-transcription and translation, omics

data, including metabolomics, may be key to revealing growth bottlenecks (Marcellin et al. 2016;

Mahamkali et al. 2020; Heffernan et al. 2022). Moreover, in contrast to batch datasets, steady-state

chemostat cultures offer better reproducibility across experiments, allowing greater insight into

process optimisation (Heffernan et al. 2020).

To understand formatotrophic growth and limitations, in this study, we investigated steady-state C_1

liquid formatotrophic fermentation in chemostat for *E. limosum*. Based on our previous batch

observations showing distinct growth phases when grown on formate and methanol, here we

hypothesised there would be limitations to continuous formate assimilation, perhaps similar to

carbon monoxide toxicity leading to oscillatory growth observed in *Clostridium autoethanogenum*

(Mahamkali et al. 2020). We tested this using differential analysis between formatotrophic and

methylotrophic conditions. Intriguingly the differences are related to both protein expression and

thermodynamics. Whilst growth was poor, unexpectedly, acetate was not the major product.

Furthermore, we observed a biofilm formation, which would negatively impact scalable production.

Anecdotally, adaptive laboratory evolution could be used to address this negative phenotype,

particularly given some *E. limosum* strains are known to not form sticky polymers that support

biofilms (Flaiz et al. 2021).

We understand there is currently a metabolic model under curation for *E. limosum*, and therefore

differential proteomic and metabolomic data presented here may be a powerful tool to help

facilitate construction of the model (Bae et al. 2021; Fackler et al. 2021).

Methods/Experimental

Bacterial Strain, Growth Medium, and Continuous Culture Conditions

Eubacterium limosum ATCC 8486 (*E. limosum*) was subject to more than 200 generations of Adaptive

Laboratory Evolution under liquid C_1 fermentation conditions (500 mM methanol, 100 mM formate).

A single colony from this cell lineage was used in all experiments and stored as glycerol stocks at -

80°C. This strain was cultivated anaerobically at 37 °C in a chemically defined phosphate buffered

medium with liquid C_1 feedstocks (**Table 1**). The medium contained per litre: 0.5 g $MgCl_2.6H_2O$, 0.5 g

NaCl, 0.13 g $CaCl_2.2H_2O$, 0.75 g NaH_2PO_4, 2.05 g Na_2HPO_4, 0.25 g KH_2PO_4, 0.5 g KCl, 2.5 NH_4Cl, 0.017 g

$FeCl_3.6H_2O$, 0.5 g L-cysteine hydrochloride monohydrate, 1 mL of 1 g/L resazurin, 10 mL trace metal

solution (TMS), 10 mL B-vitamin solution. The TMS contained per litre: 1.5 g nitrilotriacetic acid, 3 g

$MgSO_4.7H_2O$, 0.5 g $MnSO_4.H_2O$, 1 g NaCl, 0.667 g $FeSO_4.7H_2O$, 0.2 g $CoCl_2.6H_2O$, 0.2 g $ZnSO_4.7H_2O$,

0.02 g $CuCl_2.2H_2O$, 0.014 g $Al_2(SO_4)_3.18H_2O$, 0.3 g H_3BO_3, 0.03 g $Na_2MoO_4.2H_2O$, 0.028 g

$Na_2SeO_3.5H_2O$, 0.02 g $NiCl_2.6H_2O$ and 0.02 g $Na_2WO_4.2H_2O$. The B-vitamin solution contained per

litre: 20 mg biotin, 20 mg folic acid, 10 mg pyridoxine hydrochloride, 50 mg thiamine-HCl, 50 mg

riboflavin, 50 mg nicotinic acid, 50 mg calcium pantothenate, 50 mg vitamin B_{12}, 50 mg 4-

aminobenzoic acid and 50 mg thioctic acid. The Rnf complex in *E. limosum* is Na^+ dependant, and

therefore to maintain a consistent concentration across all experiments, NaCl was supplemented to

medium which did not contain sodium formate as a substrate (*i.e.* chemostats growing on

methanol). As a more oxidised substrate is required for methanol growth, gaseous CO_2 was also

provided.

Steady-state conditions were reached in 0.5 L Multifors bioreactors (infors AG) controlled by EVE

software at a working volume of 350 mL (magnetic marine impeller agitation). The system was

equipped with peristaltic pumps, mass flow controllers (MFCs), pH and temperature sensors, and

was connected to a Hiden HPR-20-QIC mass spectrometer (Hiden Analytical) for on-line off-gas

analysis. Antifoam (Sigma 435503) was continuously added to the bioreactor using a syringe pump at

10 µL/h. Results presented here are after optical density, gas uptake/production rates, and

acid/base addition rates were constant for at least five working volumes. Three technical replicate

samples, spaced by one dilution volume, were collected per biological replicate.

Experimental analysis and Quantification

Extracellular analysis

Optical density (OD) measurements were taken at 600 nm via a UV-Vis spectrophotometer (Thermo

Fischer Scientific Genesys 10S UV-Vis Spectrophotometer, USA). A biomass formula of $C_4H_7O_2N_{0.6}$

and 0.32 gDCW/L/OD was used to convert OD to molar cell concentrations (Wood et al. 2021). Liquid

samples for extracellular metabolomic analysis were collected, filtered and stored at -20 °C. Samples

were analysed by high performance liquid chromatography using an Agilent 1200 HPLC System with

Phenomenex Rezex RHM-Monosaccharide H+ column (7.8 x 300 mm, PN: OOH-0132-KO) and guard

column (Phenomenex SecurityGuard Carbo-H, PN: AJO-4490). Analytes were eluted isocratically with

4 mM H_2SO_4 at 0.6 mL min^{-1} for 48 min, and column oven temperature of 65 °C. 30 µL of sample was

injected and monitored using a UV detector (210 nm) and RID at positive polarity and 40 °C.

Bioreactor off-gas analysis was performed by an on-line mass spectrometer, monitoring the amounts

of H_2, Ar and CO_2 at 2, 40 and 44 amu respectively. To achieve reliable off-gas analysis, a bypass line

from the feed gas bottle was used as the calibration gas for each MS-cycle (*i.e.* 'on-line' calibration).

Specific rates (mmol/gDCW/d) were calculated by accounting for the exact gas composition,

bioreactor liquid working volume, feed gas flow rate, off-gas flow rate (based on a constant flow of

inert Argon), the ideal gas molar volume, pH and CO_2 dissolution equilibrium, and the steady-state

biomass concentration.

As a quality check, gaseous samples were also taken for 'off-line' analysis using a Shimadzu 2014 GC,

equipped with a ShinCarbon packed column (ST 80/100, 2mm ID, 1/8 OD Silico, Restek). H_2 and

Argon were detected by a thermal conductivity detector (TCD), while CO_2 was measured using a

flame ionization detector (FID).

Intracellular analysis

Quantitative proteome analysis was carried out using direct data-independent acquisition mass

spectrometry approach (direct-DIA). Sampling, storage and sample preparation were performed

based on a method previously developed for autotrophic growth of *C. autoethangenum*. Briefly, 5

ODs of culture was pelleted by immediate centrifugation (16,000g for 3 minutes at 4 °C) followed by

washing with 1 mL PBS (Sigma P4417). A further round of centrifugation was used, with the

supernatant then discarded and the remaining sample stored at -20 °C. Approximately 100 µg of

protein was resuspended in 25 µL Milli-Q water and combined with 25 µL SDS lysis buffer (10% SDS,

100 mM Tris). Cells grown on formate required bead beating with a small amount of 0.1 mm

diameter glass beads to lyse. To reduce protein disulphide bonds, dithiothreitol (DTT) was added to a

final concentration of 20 mM before incubating at 70 °C for 1 hour. Iodoacetamide (IAA) was added

to a final concentration of 40 mM and incubated in the dark for 30 minutes to alkylate cysteine. The

reaction was sonicated and 2.5 µL 12% phosphoric acid was added, before combining with 165 µL S-

Trap binding buffer (90% methanol, 100 mM Tris (aq)). The sample was centrifuged for 8 minutes at

13,000 g, before adding to the S-Trap spin column, spun for 1 minute at 4000g and washed thrice

with 150 µL S-Trap binding buffer. 1 µg of trypsin in 50 µL of 50 mM ammonium bicarbonate pH 8

was added to cover the top of the protein trap, incubating in a sealed bag overnight at 37 °C.

Peptides were eluted into a collection tube with 40 µL increasing concentration of acetonitrile in

0.1% formic acid. Samples were spun dry and resuspended in 20 µL buffer A (0.1% formic acid (aq))

for injection to the mass spectrometer.

Mass spectrometry for proteomics analysis was performed using LC-MS/MS with a Thermo Fisher

Scientific UHPLC system coupled to an Exploris 480 mass spectrometer. The electrospray voltage was

2.2 kV in positive-ion mode, and the ion transfer tube temperature was 250 °C. Full MS scans were

acquired in the Orbitrap mass analyser over the range of m/z 340-1110 with a mass resolution of

120,000 (at m/z 200). The automatic gain control (AGC) target value was set at 'Standard', and the

maximum accumulation time was 'Auto' for the MS/MS. The MS/MS ions were measured in 12

windows from mass 350-470, in 36 windows from mass 465-645 and 10 windows from mass 640-

1100. Analysis of data were performed using Spectronaut against a reference proteome (UniProt ID

UP000246246), with a Q-value cutoff of 0.05 applied to differential protein expressions. Locus tags

correspond to (Song et al. 2017, 2018).

Intracellular metabolome analysis was based on the method previously developed for autotrophic

growth of *C. autoethangenum* (Mahamkali et al (2020)). Briefly, 5 ODs of culture was pelleted by

immediate centrifugation (16,000g for 3 minutes at 4 °C) followed by resuspension in chilled 50%

acetonitrile to extract intracellular metabolites. Cell debris was removed by centrifugation with the

supernatant stored at -80 °C. Samples were then freeze dried and resuspended in 90 µL 2%

acetonitrile containing 5 µM azidothymidine standard. To remove lipophilic compounds (such as

lipids, fatty acids, oil), the extract was fractionated by adding 250 uL of chloroform, before adding

410 µL MQ water, vortexing, and then separating the organic and polar phases by centrifugation.

The polar fraction was cleaned through a spin column, freeze dried and resuspended with 2%

acetonitrile. LC-MS/MS analysis was performed using a Shimadzu UHPLC System (Nexera X2)

coupled to a Shimadu 8060 triple quadrupole (QqQ) mass spectrometer following (Espinosa et al.

2020; Mahamkali et al. 2020) with modifications and additions to the scheduled multiple reaction

monitoring (sMRM) transitions. Chromatographic separation was performed on a Gemini NX-C18

column (3μm x 150mm x 2mm, PN: 00F4453B0, Phenomenex) with an ion-pairing buffer system

consist of mobile phase A: 7.5mM tributylamine (pH 4.95 with acetic acid) and mobile phase B:

acetonitrile. 5 μL and 10 μL of sample were injected to ensure measured intensities fell within the

standard curve ranges.

Cell pellets were also used for PHB content analysis, following the method of (de Souza Pinto

Lemgruber et al. 2019).

Thermodynamic and kinetic metabolic flux analysis

We evaluated reaction thermodynamic driving forces using the thermodynamic variability analysis

model presented by Mahamkali et al. (2020). pH was assumed as 1 unit higher than the extracellular

pH (Lindley et al. 1987). Reaction concentrations were constrained by measured values from

intracellular metabolomics (or the lower limit of quantification, LLOQ). Carboxylate products (*i.e.*,

acetate, butyrate *etc.*) intracellular concentrations were limited according to between *ca.* 1 times

higher, or 10^6 times lower, than extracellularly, depending on possible transport mechanism (Lindley

et al. 1987; Mahamkali et al. 2020). For formate substrate intracellular concentration, this was

instead between 10 times higher, or 50 times lower, than extracellularly (Refer to Dataset A).

Beyond that, minimum concentrations were set to be 0.1 μM for metabolites, and 1 nM for

dissolved gases unless directly measured or evaluated from Henry's law (H_2, CO_2 and CO). CO_2 total

concentration, *c*, was calculated from partial pressure, *p*, as,

$$c_{co2} = p_{CO2} \times \frac{25.2mM}{atm} \times (1 + 10^{pH-6.116}) \tag{1}$$

Maximum values were set to 1 mM for activated metabolites, and 10 mM for others.

A high-level estimate for reaction fluxes was determined as follows to estimate potential bottlenecks

in metabolism. Considering metabolic flux (*J*) is controlled by protein concentration (*E*), Gibbs free

energy of reaction (Δ*G*), saturation by substrates (*M*) with affinity (*K*), kinetic orders (*a*), and other

sources of regulation (*U*), according to (Heffernan et al. 2022),

$$J \sim E \times \left(1 - e^{\Delta G/RT}\right) \times \prod_i (M_i/K_i)^{a_i} \times U(M) \tag{2}$$

The enzyme and kinetics terms can be further simplified using Michaelis-Menten enzyme kinetics for

a given substrate concentration (S) to give flux (J) as,

$$J \sim E \times k_{cat} \frac{S}{K_m + S} \times \left(1 - e^{\Delta G/RT}\right) \tag{3}$$

We can then make several assumptions to get likely order of magnitude estimates for J as follows.

• Our proteomic analysis can not determine absolute quantifications due to differences in

mass spectrometer protein constituent signal response. However using a spiked protein

analysis, (Valgepea et al. 2022) found a linear relationship between the $\log_{10}$ of mass

spectrometer intensity and $\log_{10}$ anchor protein concentrations using a similar instrument

setup to here. Since we can not be certain of the injected sample protein mass, we instead

scaled protein concentrations by the pta enzyme, which Valgepea et al. (2022) showed was

largely consistent across different gas fermentation conditions.

• Few *E. limosum* proteins have been assayed, and therefore instead we undertook BLAST

analysis of key proteins against enzymes with known kinetic parameters, k_{cat} and K_m. We

then used this as verification data against deep learning models, where protein amino acid

sequences were used to yield a complete set of kinetic parameters for the WLP central

carbon metabolism (Kroll et al. 2021; Li et al. 2021). Where intracellular substrate

concentrations, S, were unknown, instead k_{cat}/K_m was taken as a proxy for kinetic effects.

Results

Steady state fermentation

Formatotrophic grown cells reached steady-state at a dilution rate (D) of 0.4 d^{-1} (specific growth rate

of 0.017 h^{-1}). The dilution rate was chosen as the highest achievable across formatotrophic and

methylotrophic growth conditions, based on batch bottle growth rates. Despite the same amount of

carbon being supplied to each condition, formate reached only a biomass concentration of

0.07±0.02 gDCW/L, less than 15% of methylotrophic growth. We note however, that there was

significant biofilm formation in the formate condition, and so biomass concentration as reported

here may be underestimated. As a result, substrate-specific uptake rates were much higher for the

formate condition but potentially overestimated: 280±56 mmol/gDCW/d compared to 75±7.0

mmol/gDCW/d for methanol in the methanol/CO_2 condition (**Figure 1A**).

Formatotrophic growth could only be achieved when no CO_2 gas was supplied to the culture.

Further, a formate titre of 2.0±0.18 g/L was observed at steady-state (**Table 1**), indicating the culture

was not formate limited, but rather constrained by another parameter.

As expected, formate-grown cells showed significant CO_2 evolution (150±11 mmol/gDCW/d) (a

formate consumption to CO_2 production ratio of 1.9±0.4 mol/mol). However, unexpectedly, the

major non-CO_2 product under formatotrophic growth was not acetate. Only trace amounts of

acetate were detected throughout the entire fermentation (0.92±0.35 mmol/gDCW/d) (**Figure 1A**).

Other typical *E. limosum* products such as butyrate (0.32±0.13 mmol/gDCW/d), hexanoate and

butanol, were minor or below the detection limit (**Figure 1A**).

Hydrogen was produced in both cultures as a minor sink of electrons. Intriguingly, despite formate

being the less reduced substrate, it produced more hydrogen at 21±7.8 mmol/gDCW/d compared to

1.1±0.56 mmol/gDCW/d for methanol/CO_2 (**Figure 1A**). We noted in ref to other paper that formate

seems to trigger hydrogen production, which is further reinforced with this data.

Overall, this gave a carbon and electron balance of 63±15% and 24±6% for formatotrophic growth

(**Figure 1B**). These values suggest the production of an unknown product with a redox number of

5.0±2.8. Given none of the known products which branch from acetyl-CoA were detected (*e.g.*

acetate, butyrate, hexanoate, butanol *etc.*), we hypothesised there may be a product downstream of

pyruvate. Whilst no genes are identified in the *E. limosum* genome for PHB production, given the

unknown product redox number, we analysed cell pellets for intracellular PHB content. Formate

cultures showed 0.68±0.63% wPHB/wDCW. This is sufficiently low that the effect on carbon and

electron balances is negligible. We also measured for 4-aminobutyrate (GABA) (potentially produced

via partial TCA cycle) and 3-methyl-2-oxopentanoate (potentially produced *via* isoleucine

biosynthesis) using HPLC analysis against standards. As neither of them could be found, we also

looked for intermediates along the GABA pathway (2-ketoglutarate), and 3-methyl-2-oxopentanoate

pathway (citraconate, 2-oxobutanoate). The analysis confirmed that none of these compounds were

significantly detected in the samples, although the extracellular 2-ketoglutarate concentration was

above the LLOQ for the formatotrophic condition, yet it was not detected for methylotrophic growth

(data not shown).

Intracellular conditions

Given the lack of success in identifying the missing product, we performed proteomics and

metabolomics analyses to gain an understanding of formatotrophic *E. limosum* metabolism, and

compared it to methanol metabolism (ref to other paper). We identified 934 proteins that met the

criteria for differential analysis between the formate and methanol/CO_2 growth conditions (*i.e.*

present in both samples and Q-value < 0.05). Proteome analysis detected all the key enzymes of the

WLP, energy conservation, direct acetyl-CoA condensation pathway, carboxylate/alcohol production,

central carbon metabolism, shikimate pathway, partial TCA cycle, isoleucine biosynthesis, and

biofilm switches (**Figure 2** and **Figure 3**). Across these pathways, 31 intracellular metabolites were

identified in both conditions. Coenzyme A was below the lower limit of quantification (LLOQ) for all

conditions.

Formate is assimilated via the formate dehydrogenase elim_c2470 and FTHFS elim_c0957 which

have log_2 fold changes of 1.7 and 2.1 and are respectively the 66th and 8th most 'abundant' proteins.

Given the higher specific flux through the WLP for formate, it is not surprising there is upregulation

generally (elim_c0959, 1648, 1651, 1655). That said, the log_2 fold changes are mostly small and less

than 2. The WLP proteins which are more regulated in the methanol/CO_2 condition mostly relate to

the thermodynamic difficulty in reversing methylene reductase (elim_c0960-63). Several of the WLP

acs/codh complex were the most abundant in the formatotrophic proteome – elim_c1655, 1651.

Yet, acetyl-CoA intracellular concentration was one order of magnitude lower (0.014±0.003

µmol/gDCW compared to 0.300±0.133 µmol/gDCW). For reference, *C. autoethanogenum* cells under

syngas growth had an acetyl-CoA concentration of 0.358±0.140 µmol/gDCW (Mahamkali et al.

2020).

Strangely, mtaB (elim_c3293), part of the methyltransferase operon for methanol assimilation into

the WLP was also one of the most 'abundant' proteins in the formatotrophic proteome. Song and

colleagues showed no transcriptome fold change in mtaB between heterotrophic and autotrophic

(H_2/CO_2) growth, and low relative expression, suggesting that our observation here shows

formatotrophic and methylotrophic growth may share features not seen in other autotrophic

conditions (Song et al. 2018).

When it comes to energy conservation, there was little change in etf proteins (elim_c0229, 0230),

nor Rnf proteins detected (elim_c3882, 3884, 1224), yet ATPase was strongly downregulated in the

formate condition (elim_c3452, 3454, 3455, 3670) with log_2 fold changes greater than 2. ATP/ADP

intracellular ratios were relatively consistent at 1.4±0.07 for the formate condition, and 1.3±0.11 for

methanol/CO_2, noting however the ATP/ADP/AMP pool size was one order of magnitude higher in

the latter case. The sporomusa type Nfn complex was upregulated in the formate condition

(elim_c1993 (Kremp et al. 2020)) with a log_2 fold change of 2. The NADPH/NADP ratio was 0.40±0.25

and 1.0±0.056 for formate and methanol/CO_2 respectively, with the same order of magnitude total

pool size. The NADH/NAD ratios were 0.035 and 0.0011±0.0006 respectively, however the total pool

size for methanol/CO_2 was one order of magnitude higher. The formate condition ratio is similar to a

typical syngas fermentation at NADH/NAD of 0.019 and NADPH/NADP of 0.276 (Mahamkali et al.

2020), which probably reflects formate and syngas operate the WLP and Nfn in the same direction,

unlike methanol.

Considering pathways to produce carboxylates from acetyl-CoA (elim_c0231-0234, 2806, 2339)

there is mostly little change in expression ($\log_2$ fold change < 1), reinforcing that this metabolism is

regulated post-translation as acetate and butyrate were not major products under formatotrophic

growth. Both acetoacetyl-CoA and 3-hydroxybutyryl-CoA were not detected, supporting the

observation of no butyrate production. *E. limosum* codes for two different pta enzymes, one of

which is significantly upregulated under methylotrophic growth (elim_c1218; $\log_2$ fold change 2.4),

and the other significantly upregulated under formatotrophic growth (elim_c1884; $\log_2$ fold change

2.7). Despite the upregulation of elim_c1884, no acetyl-phosphate was detected for the formate

condition, yet 0.294±0.177µmol/gDCW was detected under methanol/CO_2 growth. This probably

reflects the difference in acetyl-CoA/CoA driving force. Whilst cells did not produce alcohols under

any condition, it is interesting to note identification of acetylating aldehyde dehydrogenase

(elim_c1223), AOR (elim_c3128, 0919) and alcohol dehydrogenase (elim_c2439, 2953, 0460). None

of these had $\log_2$ fold changes greater than 1.5 between the two conditions, yet it is worth noting

elim_c3128 was in the top 5% of proteome 'abundance'.

Carbon flux from the WLP is linked to pyruvate in glucogenesis via PFOR (elim_c2885) or

alternatively Pfl (elim_c0889), the latter of which is upregulated with a $\log_2$ fold change of 5.4 for

formatotrophic growth. This is the 9th most upregulated formatotrophic gene indicating that cells

may utilise formate directly to synthesise pyruvate, rather than oxidise it to generate ferredoxin and

CO_2 for use by PFOR. Pyruvate was not detected in the methanol/CO_2 condition (LLOQ 0.0016

µmol/gDCW), yet it was 0.060±0.044 µmol/gDCW in formate growth.

The remainder of the glucogenesis pathway (elim_c3055, 0452, 3997, 0104, 0106, 0107, 0421, 2869)

all have $\log_2$ fold changes less than 1.5. Two notable exceptions are TPI (elim_c0105) and FBA

(elim_1932, 2267) which are downregulated for formatotrophic growth ($\log_2$ *ca.* 2), along with parts

of the pentose phosphate pathway (elim_c0386, 0631, 1191) ($\log_2$ fold change *ca.* 2-3). Of all the

measured intermediates, only FBP was higher under the formate growth condition (yet still below

the LLOQ). This seems counterintuitive based on protein expression, unless FBP flux also goes

elsewhere for methylotrophic growth.

Branching off the glucogenesis and pentose phosphate pathways, the shikimate pathway proteins all

have $\log_2$ fold changes less than 1.5, with the sole exception of aroKL (elim_c0110) being

downregulated in formatotrophic growth with a $\log_2$ fold change of 1.8. PEP and E4P, which enter

the shikimate pathway, have lower concentrations for the formate condition, compared to

methylotrophic growth (0.028±0.009 µmol/gDCW v. 0.060±0.018 µmol/gDCW, and N.D. v

0.058±0.012 µmol/gDCW). Downstream of the shikimate pathway, anthranilate was not detected in

the former condition, and was 0.002±0.0004 µmol/gDCW in the latter case.

The partial TCA cycle, and isoleucine biosynthesis pathways both have three of the most upregulated

proteins under formatotrophic growth (11th, 28th and 43rd). GAD (elim_c2464) and its antiporter

(elim_c2465) have $\log_2$ fold changes of 5.0 and 3.6 respectively, suggesting 4-aminobutyrate (GABA)

may be an extracellular metabolite. Similarly, LeuDH (elim_c1793) has a $\log_2$ fold change of 3.9,

suggesting 3-methyl-2-oxopentanoate may also be an extracellular metabolite. ACN (elim_c1420),

which reversibly converts citrate and isocitrate was upregulated in the formate condition by a $\log_2$

fold change of 1.7. Yet, intracellular metabolite concentrations were the same in the two conditions

(0.127±0.027 µmol/gDCW v. 0.136±0.026 µmol/gDCW, and 0.093±0.017 µmol/gDCW v 0.103±0.022

µmol/gDCW). Interestingly FUM (elim_c3042, 3043) were both downregulated with a $\log_2$ fold

change of 1.4 and 2.0 respectively, which seems to correlate with the substrate, malate, having a

higher intracellular concentration for formate (0.186±0.077 µmol/gDCW) compared to

methanol/CO_2 (0.056±0.018 µmol/gDCW). Isoleucine biosynthesis cimA (elim_c3964) and leuCD

(elim_c3962) were downregulated by $\log_2$ fold changes of 4.3 and 2.6 respectively. The remainder of

the partial TCA cycle and isoleucine biosynthesis pathways showed no significant change in

regulation.

We found several significantly 'abundant' proteins that have not been annotated. These include, for

example, elim_c2694 (log$_2$ fold change 7.8, 22nd most 'abundant'), yet we note that it had no

significant transcriptome fold changes (log$_2$ < 1.5) between heterotrophic and autotrophic growth,

indicating it may be specific to C$_1$ liquid metabolism (Song et al. 2018). Further, there are also

upregulated proteins for which the annotation is ambiguous. For example, elim_c2880 is the most

upregulated protein with a log$_2$ fold change of 10.4, as well as being the 5th most 'abundant' protein.

It is annotated as a BadM/Rrf2 transcriptional regulator, associated with benzoyl-CoA degradation,

even though benzoate was not present in the medium. We do note BadL, used in PABA metabolism,

is part of this operon although not annotated in *E. limosum* (Vandrisse and Escalante-Semerena

2018). Therefore, this could *potentially* mean there is a PABA nutrient limitation (noting that PABA

was not detected intracellularly in either condition). Similarly, elim_c1937 is a folate transport

upregulated for formatotrophic growth (log$_2$ fold change of 3.8 compared to methylotrophic

growth), indicating a potential THF limitation.

Discussion

Metabolic model and redox balances

Thus far we have presented 'omics data for formatotrophic *E. limosum* cultures, with differential

analysis against methylotrophic growth. Whilst valuable conclusions can be drawn, the data takes on

further meaning when integrated together in a complete kinetic model. Since such a model does not

exist for *E. limosum*, we have instead undertaken thermodynamic calculations and various kinetics

assumptions to identify several potential metabolic bottlenecks (refer to materials and methods).

Unlike NADH/NAD and NADPH/NADP, we could not measure ferredoxin. However, upper and lower

bounds on the ferredoxin state could be calculated. From the Stn, Fd^{2-}/Fd must be > 0.075. Using

Henry's law, the dissolved hydrogen must be *ca.* 0.6 µM and so the hydrogenase suggests Fd^{2-}/Fd

must be > 2.4. With the Rnf/ATPase complex operating in the forward direction for formate

metabolism (*i.e.*, generating ATP), Fd^{2-}/Fd should be > 1.5. For PFOR to generate pyruvate, and

assuming a CoA concentration of <43 µM (LLOQ), Fd^{2-}/Fd must be <60. However, considering the

predicted kinetic parameters, it is likely that the Fd^{2-}/Fd ratio is on the lower end (*ca.* 4) to generate

sufficient flux through the mthfr, codh and fdh. However, given that ferredoxin is mainly used in the

WLP and pyruvate synthesis, this wide range of possible values ($2.4 < Fd^{2-}/Fd < 60$) does not provide

further clues as to identification of the unknown product.

Figure 4 illustrates how methanol and formate metabolism differ in the WLP by reversing key

enzymes. The hydrogenase thermodynamic driving force is much larger under formatotrophic

growth, which may explain why we observed higher hydrogen evolution. The methanol condition

has an acetyl-CoA/CoA ratio three orders of magnitude higher than the formate condition, yet we

calculated a similar thermodynamic driving force. This is because, in the latter condition, a much

lower acetate concentration is possible, hence why it is in fact possible for pta to reverse under

formatotrophic growth. We calculated this acetate limit to be in the *ca.* 1mM range (using

eQuilibrator (Flamholz et al. 2012) with acetyl-P = 12.5µM, ATP/ADP = 1.5, pH = 7.9, pMg = 3.0, ionic

strength = 0.25). This confirms, that despite significant upregulation of the pta (elim_c1884), acetate

production is limited by post-translation effects. Additionally, we cannot identify how thl is feasible

to satisfy butyrate flux, without the acetoacetyl-CoA concentration being below our assumed

minimum value of 0.1 µM, which would surely impose kinetic limitations. Thus, the model confirms

our suspicion that acetyl-CoA must be diverted elsewhere (*i.e.,* towards pyruvate).

Interestingly, the upregulated Pfl is probably the major source of pyruvate, considering

thermodynamics (**Figure 4**), protein abundance and k_{cat}/K_m (**Table 2**). Yet, the model indicates PFOR

may in fact operate in the pyruvate consuming direction during formatotrophic growth, causing a

futile metabolic cycling of pyruvate. Considering thermodynamics, protein 'abundance' and k_{cat}/K_m

(**Table 2, Figure 4**), these reinforce a mix of pre- and post-translation effects control metabolism and

may explain the lack of production of GABA (no feasible export) and 3-methyl-2-oxopentanoate

(kinetic bottleneck).

Given the Stn expression, NADPH likely has greater flux in the formate condition than methanol. This

is interesting because the NADPH/NADP is lower for formate. This could be a clue for product

identification, as production may be tied to the need to regenerate NADP. However, it is noted this

would come at an ATP cost, as ferredoxin is diverted away from Rnf/ATPase chemiosmotic ATP

generation, instead towards Stn to generate NADPH.

Cellular stress and pyruvate

Formate metabolism in acetogens appears partially controlled by protein expression in response to

cellular state. For example, since formatotrophic growth compared to methanol/CO_2 has a much

lower CO_2 driving force, it makes sense that most pyruvate flux would be through heavily

upregulated Pfl, despite PFOR being the normal mechanism for pyruvate synthesis (Neuendorf et al.

2021). Given the low acetyl-CoA/pyruvate ratio, Pfl would require a high formate concentration as

driving force, which may explain why a formate titre existed at steady-state. Previously, researchers

have suggested an increased pyruvate concentration is regulated by PPDK activity and ATP

availability, causing a drop in glucogenic intermediates for *E. limosum* (Lebloas et al. 1996). Similarly,

we found that formatotrophic growth had lower glucogenic intermediates and no change in PPDK

levels (elim_c3055) compared to methylotrophic growth. Specifically, we can see that the PEP /

pyruvate ratio is respectively 1.3±0.61 and >39±12 (adopting LLOQ for pyruvate where not

detected). This means PPDK activity is being limited by ATP availability, and so increasing pyruvate

concentration and hence the required formate driving force. This associated energy shortage allows

redirection of carbon (Lebloas et al. 1996).

We can see this redirection of carbon in the lack of acetate production, and futile upregulation of

the pta enzyme. Given that metabolites on the butyrate production pathway were not detected, this

supports the idea that a significant amount of carbon goes through pyruvate.

Of all measured intracellular metabolites, most were relatively similar for both methanol and

formate conditions. However, several which were much lower or below LLOQ for formate only.

Some of these are on the glucogenesis pathway, which is controlled by ATP availability discussed

above. We found that pools of NAD, ATP and acetyl-CoA to be one order of magnitude lower for

formatotrophic growth. This is also significantly lower than typical acetogens, despite the

NADH/NAD, ATP/ADP ratios being conserved relative to other conditions (Mahamkali et al. 2020).

Clearly, these ratios are tightly controlled, despite the lower ATP gain from growth, but the overall

pool size is not.

It is hard to understand why ATP, NAD and CoA pool sizes are smaller for formatotrophic growth

compared to methylotrophic growth based on the proteomics data given that two proteins involved

in NAD biosynthesis (elim_c1173, 3031) were upregulated. The remaining two (elim_c1197, 1880)

were not detected. This includes elim_c1880, which is used to synthesise NADP from NAD, despite

NADP(H) being detected as a metabolite. In CoA biosynthesis, only one (elim_c1486) of the four

proteins were identified in the formate condition, compared to three (elim_c1486, 3523, 2435) in

the methanol condition. The lack of protein detection may correlate with the reduced ability of cells

to synthesise NAD and CoA. GTP is required for AMP synthesis (elim_c1406, 2669), and since the

GMP flux is not directed to GTP in formatotrophic growth, this may affect the ATP pool size. Some

researchers have found pool size can be reduced by cellular stress conditions, such as pH and

temperature (Chohnan et al. 1998).

450   Formate itself may be cause of stress

Carboxylate transport in acetogens is possible *via* passive diffusion, symport of anion with a proton,

transport of anion *via* a uniport and ATP-consuming ABC transporter (Mahamkali et al. 2020). Given

we are already energetically limited, we can discount the latter process. An anion uniport (*e.g.*

elim_c1361 $\log_2$ fold change of 0.7) must balance the charge with independent proton uptake, which

could be *via* an ATPase, thus assisting with the energetic requirements. This would yield a lower

intracellular formate concentration driving force (1 mM; Dataset A), however, our model predicts

the resulting intracellular formate driving force is still sufficient for cell function.

Given acid was added throughout the experiment to maintain extracellular pH, there must have

been a net alkalinisation of the cell, which may be the cause of cellular stress. This would lead to a

more negative membrane potential that might be balanced by importing protons through an

ATPase. This could theoretically improve the overall ATP gain. It is important to note *E. limosum*

ATPases are generally thought to be sodium-dependent and we also did not find a significant

upregulation of any ATPase under formatotrophic growth (Song et al. 2017; Kang et al. 2020). Whilst

there was clearly an ATP limitation, the ATP balance itself thus remains unclear.

Additionally, formate metabolism is known to lead to a loss of the acetyl-phosphate pool via formyl-

CoA (and hence ATP) (Sly and Stadtman 1963). However, we have not identified the culprit CoA-

hydrolase enzyme in the proteome. Overall, we speculate that cellular stress caused an energy

shortage, increased pyruvate concentration, which then allowed redirection of carbon.

What does AOR do?

We found high 'abundance' of AOR, with little difference between the two conditions. Previous

researchers have noted that acetogens are regulated post-translationally (Marcellin et al. 2016;

Mahamkali et al. 2020; Heffernan et al. 2022), and we believe this could be the case for AOR. Whilst

the enzyme is known to be promiscuous, AOR is generally acknowledged not to have C_1 specificity

for formate (Bertsch and Müller 2015). The same can be said for adh. The kinetic deep learning

model, however, suggests formate consumption may be possible (as k_{cat} and K_m are similar to C_2 and

C_4 substrates, but adh k_{cat} is one order of magnitude lower for C_1; data not shown), but would result

in only *ca.* 0.2 μM accumulation of methanol, which is below our limit of detection. In principle, it is

possible, which may explain our previous observation of methanol production from formate during

early growth stages, presumably when there is a highly reduced redox pool (Wood et al. 2021).

CO_2 must be maintained low to allow growth in chemostat

We found cells were unable to grow formatotrophically when 15% CO_2 gas was present in the

headspace. Growth was restored, and steady state was reached when the gas feed contained zero

CO_2. If this requirement was thermodynamics related, the bottleneck reaction would therefore

produce CO_2. All else being equal, increasing the CO_2 concentration means the formate driving force

must increase, and there must be an upper limit to intracellular formate concentration in terms of

toxicity.

Biofilm formation

Two *E. limosum* cyclic-di-GMP-I riboswitches are known to regulate virulence, motility, quorum

sensing and biofilm (Song et al. 2017). Here, under formatotrophic growth we found upregulation of

two proteins which synthesise cyclic-di-GMP from GMP – elim_c1151, 3120 (log_2 fold change of 6.2

and 3.1 respectively). Intracellular GMP concentrations were similar for both formatotrophic growth

and methylotrophic growth at 0.008±0.005 µmol/gDCW and 0.007±0.002 µmol/gDCW, respectively.

GMP can also be used for GDP and GTP synthesis and interestingly these proteins (elim_c1967,

1274) are downregulated for formatotrophic growth (log_2 fold change 1.2 and 2 respectively).

Additionally, unlike the methanol condition, no GDP or GTP were detected under formate growth,

reinforcing that GMP flux went elsewhere. Therefore, we confirm cyclic-di-GMP has a correlation

with biofilm formation strength.

We also note GlmU (elim_c1471), which produces a precursor for the adhesion of biofilms in some

bacteria, was upregulated in the formate condition (log_2 fold change of 1.1) (Burton et al. 2006).

Interestingly GlmU transcriptome shows upregulation in heterotrophic growth, compared to

autotrophic, which matches the past *E. limosum* biofilm formation observations (Song et al. 2018).

LuxR (elim_c1099) is a transcriptional regulator known to affect biofilm formation through quorum

sensing, and here it was upregulated in formatotrophic growth (log_2 fold change of 2.4) (Chen and

Xie 2011). There was little difference in one identified protein related to cell motility (twitching)

(elim_c0255).

From the 'omics data, biofilm formation in *E. limosum* formatotrophic growth is most likely caused

by cyclic-di-GMP production, adhesive compounds and/or quorum sensing. Inhibitors for adhesive

compounds produced by GlmU include N-acetyl glucosamine-1-phosphate and iodoacetamide (Chen

et al. 2015). The latter reacts with cysteine in our medium, and so is not an option here.

Anthranilate, a tryptophan metabolite, is known to reduce intracellular cyclic-di-GMP, which should

reduce biofilm formation (Li et al. 2017). Other metabolites on the tryptophan pathway, including

indole and D-tryptophan have also been shown to influence biofilm formation. However what works

for one species, can even enhance formation in other species (Li et al. 2017).

Interestingly, we did not detect anthranilate in the formate condition, but it was 0.002±0.0004

µmol/gDCW under methylotrophic growth. Anthranilate is produced from S3P, which was three

times higher in the formate condition (0.006±0.003 µmol/gDCW versus 0.002±0.0003 µmol/gDCW

for methanol). This trend suggests that there is more flux to anthranilate, but its concentration is

being kept low, perhaps to regulate intracellular cyclic-di-GMP. Hence more biofilm formed in the

formate condition.

To investigate this relationship, we supplemented the medium with 0.8 mM anthranilate in a culture

with a pre-formed biofilm. Li and colleagues noted approximately this amount of anthranilate

reduced pre-formed biofilm coverage by up to 30% for three different species of bacteria (Li et al.

2017). However, to our surprise, this concentration of anthranilate had no visual effect on the

formed biofilm (data not shown). It is possible anthranilate may only prevent biofilm formation in *E.*

limosum, which we have not tested here. Alternatively, other biofilm inhibitors may need to be tried,

such as N-acetyl glucosamine-1-phosphate to prevent adhesive compound synthesis (Chen et al.

2015).

Conclusion

We have established a baseline formatotrophic steady-state dataset for *E. limosum* covering

phenomics, proteomics and metabolomics. Cells had a high formate-specific uptake rate. However,

unexpectedly they did not produce acetate as a major product. There is evidence of a cellular energy

limitation, which resulted in an accumulation of pyruvate intracellularly, and depletion of the total

CoA and NAD pools. Together, these redirected carbon flux away from acetate, despite upregulation

of pta, and towards products downstream of pyruvate. We contend this state of cellular stress

ultimately relates to formate itself as a substrate, which cells attempt to overcome through protein

expression, for example, upregulation of Pfl. This is interesting because acetogen metabolism is

usually controlled post-translation.

Despite having high energy efficiency, there are significant challenges to overcome for *E. limosum*

formatotrophy to be a scalable technology. Firstly, limiting biofilm formation, either through

medium supplementation or adaptive laboratory evolution, will be important to reduce operational

costs. Formatotrophy also needs to overcome poor yields and limit excess CO_2 production,

necessitating another substrate such as hydrogen or methanol. Adopting either of those would

partially defeat the purpose of using formate (*i.e.,* liquid feedstock that can be directly synthesised

from CO_2, fits with existing infrastructure and has no mass transfer limitations). Throughout this

work, we have identified exciting aspects of formate metabolism that could have significance for

acetogen applications, which include more targeted products downstream of pyruvate.

Abbreviations

General: Dilution rate; D, Lower Limit of Quantification; LLOQ; Optical Density, OD; Wood-Ljungdahl Pathway, WLP

Compounds: 2PG, 2-phosphoglycerate; 3DHQ, 3-dehydroquinate; 3DHSA, 3-dehydroshikimate; 3PG, 3-phosphoglycerate; Ac-CoA, acetyl-CoA; BPG, 1,3-biphosphoglycerate; DAHP, 3-deoxy-arabino-heptulonate 7-phosphate; DHAP, Dihydroxy acetone phosphate; E4P, Erythrose 4-phosphate; EPSP, 3-enolpyruvyl-shikimate 5-phosphate; F6P, Fructose 6-hosphate; FBP, Fructose 1,6-biphosphate; G3P, Glyceraldehyde 3-phosphate; G6P, Glucose 6-phosphate; Glc, Glucose; PEP, Phosphoenolpyruvate; PRPP, 5-phosphoribosyl diphosphate; PYR, Pyruvate; R5P, Ribose 5-phosphate; RL5P, Ribulose 5-phosphate; S3P, Shikimate 3-phosphate; S7P, Sedoheptulose 7-phosphate; THF, tetrahydrofolate; X5P, Xylulose 5-phosphate

Enzymes: ALD, aldehyde dehydrogenase; ACK, acetate kinase; ACN, anonitase; ACS, acetyl-CoA synthase; ADH, alcohol dehydrogenase; AOR, aldehyde:ferredoxin oxidoreductase; aroA, DAHP synthase; aroA, 3-phosphoshikimate 1-carboxyvinyltransferase; aroB, 3-dehydroquinate synthase; aroC, chorismate synthase; aroD, 3-dehydroquinate dehydratase; aroKL, shikimate kinase; BK, butyrate kinase; cimA, citramalate synthase; CODH, CO dehydrogenase; crt, crotonase; CS, citrate synthase; ENO, enolase; Etf-Bcd, butyryl-CoA dehydrogenase; FBA, Fructose biphosphate aldose; FBP, fructose 1; FDH, formate dehydrogenase; FTHFS, formyl-THF synthetase; FUM, fumarase; GAD, glutamate decarboxylase; GAPDH, glyceraldehyde 3-phosphate dehydrogenase; GDH, glutamate dehydrogenase; GLK, Glucokinase; hbd, 3-hydroxybutyryl-CoA dehydrogenase; IDH, isocitrate dehydrogenase; ilvBN, pyruvate:2-oxobutanoate acetaldehydetransferase; ilvC, 2-aceto-2hydroxy-butanoate:NADP+ oxidoreductase; ilvD, dihydroxy acid dehydratase; ilvE, branched chain amino acid aminotransferase; KOR, oxoglutarate ferredoxin oxidoreductase; leuB, methylmalate dehydrogenase; leuCD, methylmalate hydrolase; LeuDH, leucine dehydrogenase; MDH, malate dehydrogenase; MtaABC, methanol dependent methyl transferase; MTHFC, methenyl-THF

cylcohydrolase; MTHFD, methylene-THF dehydrogenase; MTHFR, methyltransferase / methylene-

THF reductase; PC, pyruvate carboxylase; PD, pyruvate dehydrogenase; PFKA, Phosphofructose

kinase; PFL, pyruvate formate ligase; PFOR, pyruvate ferredoxin oxidoreductase; PGAM,

phosphoglycerate mutase; PGI, Glucose 6-phosphate isomerase; PGK, phosphoglycerate kinase; PK,

pyruvate kinase; PPDK, phosphoenolpyruvate synthetase; PRPS, Ribose phosphat

epyrophosphokinase; PTA, phosphotransacetylase; PTB, phosphotransbutyrylase; PTS,

phosphotransferase system; RPE, Ribulose phosphate epimerase; RPI, Ribose phosphate isomerase;

Stn, Sporomusa type Nfn transhydrogenase; TAL, Transaldolase; thl, thiolase; TKT, Transketolase;

TPI, Trisephosphate isomerase; trpE, anthranilate synthase; ydiB, shikimate dehydrogenase

Availability of data and materials

583   The datasets used and/or analysed during the current study are available from the corresponding

584   author on reasonable request.

Competing interests

586   The authors declare that they have no competing interests

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Authors' contributions

JCW designed and conducted experiments, interpreted results and was a major contributor in writing the manuscript. RAGG jointly conducted experiments. DD performed intracellular metabolomics analysis. GT performed proteomics analysis. MRP performed extracellular metabolomics and PHB analysis. EM and BV conceived experiments and provided substantiative manuscript review. All authors read and approved the final manuscript.

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Tables

Table 1 Summary of Eubacterium limosum fermentations. N = stirrer speed; BR = biological replicates; BC = biomass
concentration. Methylotrophic data in the first row is provided for comparison from ref to other paper.

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Condition	Gas	N	BR	D	pH	BC	Acetate	Butyrate
		rpm	#	Day^{-1}	Setpoint	gDCW/L	g/L	g/L
100mM Methanol	*85% Ar, 15% CO_2*	*400*	*2*	*0.4*	*6.8*	*0.54±0.05*	*1.60±0.08*	*0.99±0.05*
100 mM Formate	100% Ar	400	2	0.4	6.8	0.07±0.02	0.014±0.006	0.006±0.001

748 *Table 2 Kinetic parameters of enzymes with highest, to our knowledge, BLAST similarity to E. limosum enzymes. Deep*
749 *learning models have been used as a comparison, and for all remaining pathways.*

Enzyme	Substrate	Cofactor	K_M (µM)	V_{max} (U/mg)	Assay k_{cat}/K_M ($M^{-1}s^{-1}$)	BLAST similarity	Deep learning prediction k_{cat}/K_M ($M^{-1}s^{-1}$)
metFV	methyl-THF						4.96E+06
MTHFD	methylene-THF	NADH	2000	1650	4.46E+05	*A woodii* (86% to elim_c0959)(Ragsdale and Ljungdahl 1984)	8.91E+04
MTHFC	methenyl-THF	NADH				*C formicaceticum* (60% to elim_c0958)(Clark and Ljungdahl 1982)	1.33E+05
FTHFS	formyl-THF	ATP	6700	58.75	3.51E+04	*Clostridium cylindrosporum* (To elim_c0957)(Buttlaire 1980)	1.32E+03
Fdh	Formate	MV	39	0.74	3.34E+04	*Shewanella oneidensis* (44% to elim_c2470)(Davies 2017)	1.65E+04
codh	CO_2	Electrode	8100 (Wang et al. 2013)		4.81E+03	*Carboxydothermus hydorgenoformas* (42% to elim_c1653) (Wang et al. 2013) *Clostridium autoethanogenum* (55% to elim_c1653)(Valgepea et al. 2022)	8.23E+04
PFOR	Acetyl-CoA	Fd^{2-}, CO_2				elim_c2885	1.47E+05
Pfl	Acetyl-CoA	Formate				elim_c0889	2.63E+05
PC	pyruvate	ATP				elim_c0058	2.89E+04
CS	Acetyl-CoA					elim_c3511	2.20E+05
ACN	Citrate					elim_c1420	2.88E+03
IDH	Isocitrate	NAD				elim_c2883	2.24E+04
GDH	2-oxoglutarate	NAD(P)H	4000	32.5	3.66E+04	*P. furiosus* (49% to elim_c2504)(Robb et al. 1992)	9.88E+04
GAD	Glutamate					elim_c2464	9.85E+03
cimA	Acetyl-CoA	Pyruvate				elim_c3964	8.26E+05
leuCD	Citramalate					elim_c3963	1.34E+05

leuCD	Citraconate					elim_c3962	1.01E+07
leuB	D-erythro-3-methylmalate	NAD				elim_c3961	1.61E+05
ilvBN	2-oxobutanoate	Pyruvate				elim_c3959/ elim_c1989	7.06E+02
ilvC	2-aceto-2-hydroxybutanoate	NAD(P)H				elim_c3965	5.22E+02
ilvD	2,3-dihydroxy-3-methylpentanoate					elim_c3960	3.24E+03
ilvE	3-methyl-2-oxopentanoate	Glutamate				elim_c0806	2.97E+03
LeuDH	isoleucine	NAD				elim_c1793	7.26E+02

Figures

Figure 1 Characteristics of Eubacterium limosum in autotrophic chemostats, with conditions as summarised in Table 1.
Methylotrophic data is provided for comparison from ref to other paper. (A) specific uptake and production rates. (B)
carbon distribution of products. Values represent the average ± standard deviation between biological replicate triplicates.

Figure 2 Schematic representation of Wood Ljungdahl Pathway metabolism in E. limosum during C_1 fermentation. Reactions
are shown with substrates, products and redox mediators, however without stoichiometric balances along pathways.
Enzymes and associated protein accession numbers (elim_cxxxx) are shown for each reaction, along with log_2 fold changes
comparing methylotrophic <mark>ref to other paper</mark> and formatotrophic growth. Intracellular metabolites concentrations
(μmol/gDCW) are shown for select metabolites as the average ± standard deviation of biological replicate triplicates. The
methanol condition is in blue, and formate condition in black. The lower limit of quantification is shown as a red dashed
line, which differs between the conditions due to the differences in cell pellet mass.

Figure 3 Schematic representation of central carbon metabolism metabolism in E. limosum during C₁ fermentation.
Reactions are shown with substrates, products and redox mediators, however without stoichiometric balances along
pathways. Enzymes and associated protein accession numbers (elim_cxxxx) are shown for each reaction, along with log₂
fold changes comparing methylotrophic ref to other paper *and formatotrophic growth. Intracellular metabolites*
concentrations (μmol/gDCW) are shown for select metabolites as the average ± standard deviation of biological replicate
triplicates. The methanol condition is in blue, and formate condition in black. The lower limit of quantification is shown as a
red dashed line, which differs between the conditions due to the differences in cell pellet mass.

Figure 4 tMFA of E. limosum showing maximum allowable range of Gibbs free energy for respective reactions of the
methanol/CO_2 (blue) and formate (brown) conditions using thermodynamic variability analysis. Reaction directions are in
the anabolic direction from CO_2 on Figure 2, Figure 3. Rnf, PFOR, hydrogenase and stn are all calculated in the ferredoxin-
consuming direction, and ATP synthase in the ATP-forming direction. Pfl is not shown for methylotrophic growth as it was
not detected in the proteome.