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## Recent Work

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1 **Functional analysis of the fatty acid and alcohol metabolism of *Pseudomonas putida* using**  
2 **RB-TnSeq**

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42 **ABSTRACT**

43           With its ability to catabolize a wide variety of carbon sources and a growing engineering  
44 toolkit, *Pseudomonas putida* KT2440 is emerging as an important chassis organism for  
45 metabolic engineering. Despite advances in our understanding of this organism, many gaps  
46 remain in our knowledge of the genetic basis of its metabolic capabilities. These gaps are  
47 particularly noticeable in our understanding of both fatty acid and alcohol catabolism, where  
48 many paralogs putatively coding for similar enzymes co-exist making biochemical assignment  
49 via sequence homology difficult. To rapidly assign function to the enzymes responsible for these  
50 metabolisms, we leveraged Random Barcode Transposon Sequencing (RB-TnSeq). Global  
51 fitness analyses of transposon libraries grown on 13 fatty acids and 10 alcohols produced strong  
52 phenotypes for hundreds of genes. Fitness data from mutant pools grown on varying chain length  
53 fatty acids indicated specific enzyme substrate preferences, and enabled us to hypothesize that  
54 DUF1302/DUF1329 family proteins potentially function as esterases. From the data we also  
55 postulate catabolic routes for the two biogasoline molecules isoprenol and isopentanol, which are  
56 catabolized via leucine metabolism after initial oxidation and activation with CoA. Because fatty  
57 acids and alcohols may serve as both feedstocks or final products of metabolic engineering  
58 efforts, the fitness data presented here will help guide future genomic modifications towards  
59 higher titers, rates, and yields.

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## 65 **IMPORTANCE**

66 To engineer novel metabolic pathways into *P. putida*, a comprehensive understanding of  
67 the genetic basis of its versatile metabolism is essential. Here we provide functional evidence for  
68 the putative roles of hundreds of genes involved in the fatty acid and alcohol metabolism of this  
69 bacterium. These data provide a framework facilitating precise genetic changes to prevent  
70 product degradation and channel the flux of specific pathway intermediates as desired.

71

## 72 **INTRODUCTION**

73 *Pseudomonas putida* KT2440 is an important metabolic engineering chassis, which can  
74 readily metabolize compounds derived from lignocellulosic and plastic derived feedstocks (1–3),  
75 and has an ever-growing repertoire of advanced tools for genome modification (4–7). Its upper  
76 glycolytic pathway architecture enables *P. putida* to natively generate large amounts of reducing  
77 equivalent (8), and it more robustly withstands metabolic burdens than many other frequently  
78 used host organisms (9). To date, a wide variety of products have been produced through  
79 metabolic engineering of *P. putida*, including valerolactam (10), curcuminoids (11), diacids (12),  
80 methyl-ketones (13), rhamnolipids (14), cis,cis-muconic acid (15), and many others (16). Recent  
81 advances in genome-scale metabolic modeling of *P. putida* make engineering efforts more  
82 efficient (7, 17). However, a large gap still exists between genes predicted to encode enzymatic  
83 activity and functional data to support these assumptions. Recent characterizations of enzymes  
84 and transporters involved in the catabolism of lysine (12, 18), levulinic acid (19), and aromatic  
85 compounds (20) highlight the need to continue functionally probing the metabolic capabilities of  
86 *P. putida*, because its native catabolism can consume many target molecules and dramatically  
87 impact titers.

88           Amongst the most important metabolisms not yet rigorously interrogated via omics-level  
89 analyses are fatty acid and alcohol degradation. Recently, fatty acids have been shown to be a  
90 non-trivial component of some feedstock streams (1) and, depending on their chain length, serve  
91 as high-value target molecules (21). Furthermore, intermediates in beta-oxidation can be  
92 channeled towards mega-synthases to produce more complex molecules (22), or used in reverse  
93 beta-oxidation to produce compounds such as medium chain n-alcohols (23). However,  
94 assigning the genetic basis of fatty acid degradation is complicated by the presence of multiple  
95 homologs of the individual *fad* genes encoded in the genome of *P. putida* KT2440 (17, 24).  
96 Although work has been done to either biochemically or genetically demonstrate the substrate  
97 specificity of some individual *fad* genes, the majority of these homologs still have no functional  
98 data associated with them.

99           *P. putida* is also able to oxidize and catabolize a wide variety of alcohols. Much work has  
100 focused on the unique biochemistry and regulation of two pyrroloquinoline quinone (PQQ)-  
101 dependent alcohol dehydrogenases (ADH), *pedE* and *pedH*, which exhibit broad substrate  
102 specificity for both alcohols and aldehydes (25, 26). Specific work has also investigated the  
103 suitability of *P. putida* for the production of ethanol (27) and the genetic basis for its ability to  
104 catabolize butanol and 1,4-butanediol (28–30). *P. putida* is also known for its ability to tolerate  
105 solvents and alcohols, making it an attractive host for their industrial production (31, 32).  
106 Tolerance to these compounds is a product of both robust efflux pumps (31) and the ability of  
107 some strains, such as *P. putida* mt-2, to catabolize a range of organic compounds (33). Metabolic  
108 engineering has biologically produced a diverse range of alcohols with a wide array of industrial  
109 and commercial uses (34–36). As more alcohol synthesis pathways are engineered into *P. putida*,

110 a more complete understanding of the molecular basis of its catabolic capacities will be required  
111 to achieve high-titers.

112 A recent surge in omics-level data has revealed much about the metabolism of *P. putida*,  
113 with adaptive evolution (30), proteomics (10, 28, 29), and <sup>13</sup>C flux analysis (37–39) yielding  
114 valuable insights. An approach that has proven to be particularly powerful is Random Barcode  
115 Transposon Sequencing (RB-TnSeq) (40, 41). RB-TnSeq allows rapid and inexpensive genome-  
116 wide profiling of individual gene fitness in various conditions, and has been used in *P. putida* to  
117 identify numerous novel metabolic pathways and aid in increasing titers of the polymer precursor  
118 valerolactam (10, 11, 18–20). RB-TnSeq improves over other TnSeq-based techniques by  
119 introducing a random nucleotide barcode into the transposon which is flanked by conserved  
120 primer binding sites (40). After one initial round of TnSeq to map the transposon insertion within  
121 the genome which also associates that insertion site with a barcode, all subsequent mutant  
122 abundance quantification can be performed using standard barcode counting via Illumina  
123 sequencing of PCR product from the conserved priming sites (40). This advance reduces the cost  
124 per experiment, as well as greatly speeding up processing time required to conduct genome-wide  
125 fitness experiments (40). Here, we leverage RB-TnSeq to interrogate the genetic basis for the  
126 catabolism of multiple fatty acids and alcohols to develop an evidence-based understanding of  
127 the enzymes and pathways utilized in these metabolisms.

128

## 129 **RESULTS AND DISCUSSION**

130 Global Analysis of Fatty Acid Metabolism. To characterize the genetic basis of fatty acid  
131 metabolism in *P. putida*, barcoded transposon mutant libraries were grown in minimal media  
132 with straight chain fatty acids (C3-C10, C12, and C14), fatty esters (Tween20 and butyl stearate),

133 and an unsaturated fatty acid (oleic acid) as sole carbon sources. An overview of sample  
134 collection is provided in Figure 1A. Pearson correlation analyses of global fitness patterns  
135 revealed that the metabolisms of straight chain fatty acids between C7 and C14 clade together,  
136 suggesting similar overall catabolic routes (**Figure 1B**). Oleic acid, an 18-carbon  
137 monounsaturated fatty acid, also grouped within this clade. Shorter chain fatty acids (<C7) did  
138 not show high correlation to one another based on global fitness analyses, suggesting more  
139 independent routes of catabolism (**Figure 1B**). Annotations in the BioCyc database, functional  
140 assignment from a recent metabolic model of *P. putida* KT440 (*iJN1462*), and previous *in vitro*  
141 biochemical work predict the existence of several enzymes in the genome of the bacterium that  
142 may be putatively involved in fatty acid catabolism: six acyl-CoA ligases, seven acyl-CoA  
143 dehydrogenases, seven enoyl-CoA hydratases, four hydroxyacyl-CoA dehydrogenases, and five  
144 thiolases (**Figure 2**) (17, 24, 42). Our data show discrete fitness patterns for the steps of beta-  
145 oxidation that appear to be largely dictated by chain length (**Figure 2**).

146         When grown on fatty acids, many bacteria require the anaplerotic glyoxylate shunt to  
147 avoid depleting TCA cycle intermediates during essential biosynthetic processes. In *P. putida*,  
148 the two steps of the glyoxylate shunt are encoded by PP\_4116 (*aceA* - isocitrate lyase) and  
149 PP\_0356 (*glcB* - malate synthase). Transposon mutants in both of these genes showed serious  
150 fitness defects (fitness score < -3) when grown on nearly all of the fatty acids tested (**Figure 2**).  
151 However, the glyoxylate shunt genes appeared dispensable for growth on valerate (C5), and  
152 showed a more severe fitness defect when grown on heptanoate (C7). Complete beta-oxidation  
153 of valerate and heptanoate results in ratios of propionyl-CoA to acetyl-CoA of 1:1 and 1:2,  
154 respectively. This higher ratio of 3-carbon to 2-carbon production presumably offers an alternate  
155 means to replenish TCA cycle intermediates in the absence of a glyoxylate shunt (**Figure 2**).

156 In order to utilize the propionyl-CoA generated by beta-oxidation of odd-chain fatty  
157 acids, bacteria often employ the methylcitrate cycle (MCC), producing succinate and pyruvate  
158 from oxaloacetate and propionyl-CoA. In *P. putida*, the MCC is catalyzed via methylcitrate  
159 synthase (*prpC* - PP\_2335), 2-methylcitrate dehydratase (*prpD* - PP\_2338, or *acnB* - PP\_2339),  
160 aconitate hydratase (*acnB* - PP\_2339, or *acnA2* - PP\_2336), and 2-methylisocitrate lyase (*mmgF*  
161 - PP\_2334) (**Supplementary Figure 1**). Unsurprisingly, the MCC appeared to be absolutely  
162 required for growth on propionate (C3), valerate (C5), heptanoate (C7), and nonanoate (C9),  
163 with PP\_2334, PP\_2335, and PP\_2337 (a putative AcnD-accessory protein) showing severe  
164 fitness defects (**Figure 2, Supplementary Figure 1**). While PP\_2338 (*prpD*) encodes for a 2-  
165 methylcitrate dehydratase, transposon mutants showed no fitness defects when grown on odd-  
166 chain fatty acids. This reaction is likely carried out by PP\_2339 (*acnB* - a bifunctional 2-  
167 methylcitrate dehydratase/acnitate hydratase B); however, there were no mapped transposon  
168 insertions for this gene (**Figure 2, Supplementary Figure 1**). This suggests that PP\_2339 was  
169 essential during the construction of the RB-TnSeq library. Furthermore, PP\_2336 showed  
170 relatively modest fitness defects when grown on propionate and other odd-chain fatty acids,  
171 suggesting that PP\_2339 likely accounts for much of the methylaconitate hydratase activity as  
172 well (**Figure 2, Supplementary Figure 1**).

173

174 Long and Medium Chain Fatty Acid Catabolism. Pearson correlation analysis of fitness data  
175 indicated that both long and medium chain fatty acids are likely catabolized via similar  
176 pathways. Fitness data suggests that FadD1 (PP\_4549) catalyzes the initial CoA-ligation of C7 to  
177 C18 fatty acids, and may potentially act on C6 as well (**Figure 2**). Disruption of *fadD2*  
178 (PP\_4550) did not cause fitness defects as severe as those seen in *fadD1* mutants, although it did



179 result in moderate fitness defects when grown on C8-C10 fatty acids. These data are consistent  
180 with the biochemical characterization of FadD1 from *P. putida* CA-3, which showed greater  
181 activity on longer chain alkanolic and phenylalkanoic acids than on shorter chain substrates (43).  
182 For fatty acids with chain lengths of C10 and greater, the data suggest that the *fadE* homolog  
183 PP\_0368 is the primary acyl-coA dehydrogenase, while the nearby *fadE* homolog PP\_0370  
184 appears to be preferred for C6-C8 fatty acids (**Figure 2**). A relatively even fitness defect for  
185 these two *fadE* homologs indicates that PP\_0368 and PP\_0370 may have equal activity on  
186 nonanoate (**Figure 2**). These data are supported by a previous biochemical characterization of  
187 PP\_0368, in which it showed greater activity on chain lengths longer than C9 (44). The *fadB*  
188 homolog PP\_2136 showed severe fitness defects when grown on all fatty acids with chain  
189 lengths of C6 and longer, implicating it as the primary enoyl-CoA hydratase/3-hydroxy-CoA  
190 dehydrogenase for those substrates (**Figure 2**). *P. putida* was able to grow on the unsaturated  
191 substrate oleic acid, and is likely able to isomerize the position of the unsaturated bond via the  
192 enoyl-CoA isomerase PP\_1845, which showed specific fitness defects when grown on oleic acid  
193 (**Figure 2**). *P. putida*'s primary long chain thiolase appears to be the *fadA* homolog PP\_2137,  
194 which showed severe to moderate fitness defects when grown on fatty acids with chain lengths  
195 C8 or longer (**Figure 2**). Fitness data for mutant pools grown on heptanoate showed minor  
196 fitness defects for both PP\_2137 and PP\_3754 (*bktB*), suggesting that both thiolases may work  
197 on C7 substrates (**Figure 2**).

198 Both long chain fatty esters tested (Tween 20 and butyl stearate) appeared to utilize the  
199 same *fad* homologs as the long chain fatty acids. However, before either molecule can be  
200 directed towards beta-oxidation, Tween 20 and butyl stearate must be hydrolyzed to generate a  
201 C12 or C18 fatty acid, respectively. To date, no such hydrolase has been identified in *P. putida*

202 KT2440. Comparing the mutant fitness scores between Tween 20 and laurate (C12) carbon  
203 source experiments revealed six genes (PP\_0765, PP\_0766, PP\_0767, PP\_0914, PP\_2018, and  
204 PP\_2019) that had significant and severe fitness defects specific to Tween 20 (fitness score  $< -2$ ,  
205  $t > |4|$ ) in both biological replicates (**Figure S2**). The same comparison between butyl stearate  
206 and myristate (C14) revealed four genes specific to the fatty ester (PP\_0765, PP\_0766, PP\_2018,  
207 and PP\_4058) that had significant severe fitness defects (fitness score  $< -2$ ,  $t > |4|$ ) in both  
208 biological replicates (**Figure S2**). As PP\_0765-6 and PP\_2018 appear to have specific  
209 importance in both of the ester conditions tested, it may be possible that they contribute to the  
210 hydrolysis of the fatty ester bonds. However, it is also possible that the esterase is secreted or  
211 associated with the outer membrane (45), in which case its enzymatic activity would be shared  
212 amongst the library and it would not have the associated fitness defect expected (10).

213         The genes PP\_2018 and PP\_2019 encode a BNR-domain containing protein and a RND-  
214 family efflux transporter, respectively, and are likely co-expressed in an operon that also  
215 includes PP\_2020 and PP\_2021. Interestingly, although PP\_2021 codes for a putative lactonase,  
216 transposon mutants had no apparent fitness defect with either of the fatty esters as the carbon  
217 source. PP\_0765 and PP\_0766 encode a DUF1302 family protein and DUF1329 family protein,  
218 respectively. Given their similar fitness scores, they are likely coexpressed in an operon  
219 positively regulated by the LuxR-type regulator PP\_0767 (**Figure S2**). Previous work in multiple  
220 other species of *Pseudomonas* has observed cofitness of DUF1302/DUF1329 family genes with  
221 BNR-domain and RND-family efflux genes when grown on Tween 20 (41). The authors  
222 proposed that these genes may work together in order to export a component of the cell wall.  
223 However, an alternative hypothesis could be that PP\_0765 and PP\_0766 contribute to catalyzing  
224 the hydrolysis of fatty esters, accounting for the missing catabolic step of butyl stearate and

225 Tween 20. This hypothesis is bolstered somewhat by the co-localization of PP\_0765/PP\_0766  
226 near fatty acid catabolic genes in *P. putida* KT2440 and many other *Pseudomonads* (**Figure S3**).  
227 Future work will need to be done to biochemically characterize the potential enzymatic activity  
228 of these proteins.

229

### 230 Short Chain Fatty Acid Catabolism.

231 In our genome-wide fitness assays, the mutant fitness patterns of C6 or shorter fatty acid  
232 carbon sources had lower Pearson correlation between one another than the correlations within  
233 long and medium-chain fatty acids (**Figure 1**). These global differences reflect what appear to be  
234 discrete preferences in beta-oxidation enzymes for growth on short chain fatty acids. Fitness data  
235 suggest that while both CoA-ligases PP\_0763 and PP\_4559 are required for growth on  
236 hexanoate, only PP\_0763 is required for growth on valerate (**Figure 2**). Furthermore, the  
237 putative positive regulator of PP\_0763, LuxR-family transcription factor PP\_0767, also showed  
238 a significant fitness defect (-2.0) when grown on both valerate and hexanoate (**Figure 2**).  
239 PP\_0370 seems to be the acyl-CoA dehydrogenase largely responsible for hexanoate catabolism,  
240 though PP\_3554 mutants also have minor fitness defects. The dehydrogenation of valeryl-coA  
241 appears to be distributed between the activities of PP\_0368, PP\_0370, and PP\_3554, with no  
242 single acyl-CoA dehydrogenase mutant demonstrating a strong fitness defect when grown on  
243 valerate (**Figure 2**). Interestingly, though previous biochemical analysis had demonstrated that  
244 PP\_2216 has activity on C4-C8 acyl-CoA substrates with a preference for shorter chain lengths  
245 (46), we observed no fitness defects for PP\_2216 mutants when grown on any fatty acid carbon  
246 source (**Figure 2**).

247           It appears that the role of enoyl-CoA hydratase or hydroxyacyl-CoA dehydrogenase may  
248 be distributed across multiple enzymes for both hexanoate and valerate. Growth on hexanoate  
249 resulted in moderate fitness defects in mutants disrupted in the predicted enoyl-CoA hydratases  
250 PP\_2136, PP\_2217, and PP\_3726; however, for mutants grown on valerate, there were almost no  
251 observable fitness defects for any of the enoyl-CoA hydratase enzymes examined in the study,  
252 suggesting that for this chain length significant functional complementation exists between the  
253 *fadB* homologs (**Figure 2**). Fitness data suggest that PP\_2136 (*fadB*), PP\_2214 (a predicted type  
254 II 3-hydroxyacyl-CoA dehydrogenase), and PP\_3755 (a 3-hydroxybutyryl-CoA dehydrogenase)  
255 may all be involved in the dehydrogenation of 3-hydroxyhexanoyl-CoA (**Figure 2**), while there  
256 appears to be a distribution of *fadB*-like activity when it comes to the dehydrogenation of 3-  
257 hydroxyvaleryl-CoA, with PP\_3755 showing only a slight fitness defect on valerate.  
258 Intriguingly, mutants disrupted in the predicted type-2 acyl-CoA dehydrogenase PP\_2214  
259 showed apparent increased fitness when grown on valerate (**Figure 2**). As with heptanoate,  
260 fitness data from mutant pools grown on valerate or hexanoate suggest that both PP\_2137 and  
261 PP\_3754 may catalyze the terminal thiolase activity of these substrates. The lack of pronounced  
262 fitness phenotypes for the beta-oxidation steps of both valerate and hexanoate underscores the  
263 necessity for further *in vitro* biochemical interrogation of these pathways.

264           Both the butanol and butyrate metabolism of *P. putida* have been studied in detail  
265 through omics-level interrogation across multiple strains (28, 29). Previous work showed that  
266 during growth on n-butanol, which is later oxidized to butyrate, three CoA-ligases are up-  
267 regulated: PP\_0763, PP\_3553, and PP\_4487 (*acsA-1* - an acyl-CoA synthase) (29). However, our  
268 butyrate carbon source experiments only revealed strong fitness defects in PP\_3553 mutants  
269 (**Figure 2, Figure 3A**). The same work found that PP\_3554 was the only upregulated acyl-CoA

270 dehydrogenase, which agrees with the strong fitness defect we observed in mutants of that gene  
271 (29). That prior work did not find upregulation of any enoyl-CoA hydratase in *P. putida* grown  
272 on butanol, but this is likely reflective of redundancy in this step; we observed fitness defects in  
273 multiple genes, including PP\_2136, PP\_2217, and PP\_3726, with mutants in PP\_2217  
274 demonstrating the most severe fitness defect (**Figure 2, Figure 3A**). Hydroxyacyl-CoA  
275 dehydrogenase PP\_2136 and 3-hydroxybutyryl-CoA dehydrogenase PP\_3755 (*hbd*) have both  
276 been shown to be upregulated during growth on butanol (29). While our data showed fitness  
277 defects in both of these genes, the defect of PP\_3755 mutants was much more severe. Three  
278 different thiolases (PP\_2215, PP\_3754, and PP\_4636) and the 3-oxoacid CoA-transferase *atoAB*  
279 were previously observed to be upregulated during growth on butanol, but of these genes, only  
280 PP\_3754 (*bktB*) had a strong fitness defect, implying that it is the main thiolase for the terminal  
281 step of butyrate catabolism (**Figure 2, Figure 3A**).

282         The inability of the RB-TnSeq data to clearly show which enzymes are likely responsible  
283 for specific beta-oxidation reactions suggest multiple enzymes may catalyze these steps. In  
284 addition to the lack of genotype to phenotype clarity in the enzymes responsible for the catabolic  
285 steps, we observed additional phenotypes within our fitness data that portray a complex picture  
286 of short chain fatty acid metabolism in *P. putida*. The TetR-family repressor *paaX* (PP\_3286)  
287 was shown to have a negative fitness score when mutant pools were grown on fatty acids with  
288 chain lengths C7 or below (**Figure S4**). PaaX negatively regulates the *paa* gene cluster encoding  
289 the catabolic pathway for phenylalanine (47, 48), implying that presence of phenylalanine  
290 catabolism impedes growth on short chain fatty acids. It is therefore somewhat surprising that no  
291 individual mutant within the *paa* gene cluster shows a fitness increase when grown on short

292 chain fatty acids, though no robust fitness data exists for *paaJ* (PP\_3275 - a 3-oxo-5,6-  
293 didehydrosuberil-coA thiolase) (**Figure S4**).

294 Mutants in MerR-family regulator PP\_3539 showed very high fitness benefits (fitness  
295 scores of 3.8 and 4.7 in two biological replicates) when grown on valerate. PP\_3539 likely  
296 increases expression of *mvaB* (PP\_3540 - hydroxymethyl-glutaryl-CoA lyase), thus suggesting  
297 that decreased levels of MvaB activity may benefit *P. putida* valerate catabolism. Unfortunately,  
298 there are no fitness data available for *mvaB*, likely because it is essential under the conditions in  
299 which the initial transposon library was constructed. The genes *hdb* and *bktB*, encoding the  
300 terminal two steps of butyrate metabolism, are flanked upstream by an AraC-family regulator  
301 (PP\_3753) and downstream by a TetR-family regulator (PP\_3756); the latter is likely co-  
302 transcribed with the butyrate catabolic genes (**Figure 3B**). When grown on butyrate, mutants in  
303 both PP\_3753 and PP\_3756 show decreased fitness; however, previous work to evaluate global  
304 fitness of *P. putida* grown on levulinic acid showed negative fitness values only for PP\_3753,  
305 *htb*, and *btkB* (**Figure 3B**). These results suggest that the TetR repressor may be responding to a  
306 butyrate specific metabolite. Finally, across multiple fitness experiments, the TonB siderophore  
307 receptor PP\_4994 and the TolQ siderophore transporter PP\_1898 showed fitness advantages  
308 when grown on fatty acids, especially on hexanoate (**Figure S5**). Together, these results suggest  
309 that a wide range of environmental signals impact how *P. putida* is able to metabolize short  
310 chain fatty acids.

311

### 312 Global Analysis of Alcohol Catabolism

313 In addition to its ability to robustly catabolize a wide range of fatty acid substrates, *P.*  
314 *putida* is also capable of oxidizing and catabolizing a wide variety of alcohols into central

315 metabolism through distinct pathways. To further our understanding of these pathways,  
316 transposon libraries were grown on a number of short n-alcohols (ethanol, butanol, and  
317 pentanol), diols (1,2-propanediol, 1,3-butanediol, 1,4-butanediol, and 1,5-pentanediol), and  
318 branched chain alcohols (isopentanol, isoprenol, and 2-methyl-1-butanol). Relative to growth on  
319 fatty acids, fitness experiments of *P. putida* grown on various alcohols showed less correlation to  
320 one another, reflecting the more diverse metabolic pathways used for their catabolism (**Figure**  
321 **4A**). The initial step of the catabolism of many primary alcohols is the oxidation of the alcohol to  
322 its corresponding carboxylic acid. The BioCyc database features 14 genes annotated as alcohol  
323 dehydrogenases (PP\_1720, PP\_1816, PP\_2049, PP\_2492, PP\_2674, PP\_2679, PP\_2682,  
324 PP\_2827, PP\_2953, PP\_2962, PP\_2988, PP\_3839, PP\_4760, and PP\_5210) (24). Fitness data  
325 showed that the majority of these alcohol dehydrogenases had no fitness defects when grown on  
326 the alcohols used in this study (**Figure 4B**).

327         The alcohol dehydrogenases that showed the most consistent fitness defects across  
328 multiple conditions were the two PQQ-dependent alcohol dehydrogenases PP\_2674 (*pedE*) and  
329 PP\_2679 (*pedH*), as well as the Fe-dependent alcohol dehydrogenase PP\_2682 (*viaY*) (**Figure**  
330 **4B**). Both *pedE* and *pedH* have been extensively studied in *P. putida* and other related bacteria,  
331 and are known to have broad substrate specificities for alcohols and aldehydes (25, 26, 49). Their  
332 activity is dependent on the activity of *pedF* (PP\_2675), a cytochrome *c* oxidase that regenerates  
333 the PQQ cofactor (25). In *P. aeruginosa*, a homolog of *viaY* (*ercA*) was shown to have a  
334 regulatory role in the expression of the *ped* cluster, and was not believed to play a direct  
335 catabolic role (50). Recent work has validated that this function is conserved in *P. putida* (51). In  
336 most conditions tested, disruption of *pedF* caused more severe fitness defects than either *pedE* or  
337 *pedH* individually, suggesting they can functionally complement one another in many cases.

338 However, growth on 2-methyl-1-butanol and 1,5-pentanediol both showed more severe fitness  
339 defects in *pedE* mutants compared to *pedF* (**Figure 4B**). In many other alcohols, including  
340 ethanol and butanol, even disruption of *pedF* did not cause extreme fitness defects, suggesting  
341 the presence of other dehydrogenases able to catalyze the oxidation (**Figure 4B**).

342 The transcriptional regulatory systems that activate expression of various genes in the *ped*  
343 cluster could also be identified from these data. Mutants in either member of the sensory  
344 histidine kinase/response regulator (HK/RR) two component system, *pedS2/pedR2*, showed  
345 significant fitness defects when 2-methyl-1-butanol was supplied as the sole carbon source. This  
346 HK/RR signaling system has been shown to activate the transcription of *pedE* and repress *pedH*  
347 in the absence of lanthanide ions (52). Since lanthanides were not supplied in the medium, this  
348 likely explains the fitness defect observed in *pedS2/pedR2*. The transcription factor *pedR1*  
349 (*agmR*) was also found to affect host fitness when grown on various alcohols (**Figure 5**). This  
350 gene has been identified in *P. putida* U as an activator of long chain (C6+) n-alcohol and  
351 phenylethanol catabolism (53). In *P. putida* KT2440, *pedR1* has been associated with the host  
352 response to chloramphenicol, and its regulon has been elucidated previously (54). Our data  
353 reflect the literature, indicating that *pedR1* functions as a transcriptional activator of the *ped*  
354 cluster and *pedR2* functions as a specific regulator of *pedE* and *pedH*.

355 Unsurprisingly, the genes required for the biosynthesis of the PQQ cofactor were also  
356 amongst the most co-fit (cofitness is defined as Pearson correlation between fitness scores of two  
357 genes over many independent experimental conditions) with both *pedF* and *yiaY*. *P. putida*  
358 synthesizes PQQ via a well-characterized pathway, starting with a peptide encoded by the gene  
359 *pqqA* (PP\_0380) which is then processed by *pqqE*, *pqqF*, and *pqqC* to generate the final cofactor  
360 (**Figure 4C**). The three synthetic genes (*pqqEFC*) all showed significant fitness defects on the



361 same alcohols as the *pedF* mutants, while *pqqA* showed a less severe fitness phenotype (**Figure**  
362 **4C**). However, the small size of *pqqA* resulted in few transposon insertions, making it difficult to  
363 draw confident conclusions. Two genes showed similar defective fitness patterns on select  
364 alcohols: *pqqB*, which has been proposed to be an oxidoreductase involved in PQQ biosynthesis;  
365 and *pqqD*, a putative PQQ carrier protein. Previous work regarding a PqqG homolog from  
366 *Methylobacterium extorquens* suggested that it forms a heterodimeric complex with PqqF that  
367 proteolytically processes PqqA peptides, although PqqF was sufficient to degrade PqqA on its  
368 own (55). Fitness data from *P. putida* may support this hypothesis, as there was no observed  
369 fitness defect in *pqqG* mutants when grown on any alcohol, suggesting that the bacterium is still  
370 able to process PqqA with PqqF alone (**Figure 4C**).

371

#### 372 Short chain alcohol metabolism

373 The metabolism of n-alcohols almost certainly proceeds through beta-oxidation using the  
374 same enzymatic complement as their fatty acid counterparts. This relationship is reflected in the  
375 high correlation in global fitness data between alcohols and fatty acids of the same chain length  
376 (ethanol and acetate -  $r = 0.72$ , butanol and butyrate -  $r = 0.66$ , pentanol and valerate -  $r = 0.72$ ).  
377 However, given previous work and our fitness data, the initial oxidation of these alcohols  
378 appears to be quite complex. Biochemical characterization of both PedE and PedH have shown  
379 that both have activity on ethanol, acetaldehyde, butanol, butyraldehyde, hexanol, and  
380 hexaldehyde (25). When grown on n-pentanol, mutants disrupted in *pedF* show severe fitness  
381 defects, suggesting that PedH and PedE are the primary dehydrogenases responsible for pentanol  
382 oxidation (**Figure 4B**, **Figure 5A**). However, when grown on either ethanol or n-butanol, both  
383 the PQQ-dependent alcohol dehydrogenases (PQQ-ADHs) and *pedF* show less severe fitness

384 defects compared to when they are grown on pentanol (**Figure 4B**). This implies that other  
385 dehydrogenases are also capable of these oxidations. One likely candidate may be PP\_3839,  
386 which shows a minor fitness defect when grown on n-butanol and has been biochemically shown  
387 to oxidize coniferyl alcohol (**Figure 4B**) (56). Individual gene deletion mutants of either *pedF*  
388 (PP\_2675) or PP\_3839 showed only minor growth defects when grown on either ethanol,  
389 butanol, or pentanol as a sole carbon source (**Figure 7**). However, when both genes were deleted,  
390 no growth was observed on these substrates, suggesting that the PQQ-ADHs and PP\_3839 are  
391 the primary dehydrogenases responsible for the oxidation of short chain n-alcohols (**Figure 7**).

392         It is ambiguous from our data and from previous work which enzymes are oxidizing the  
393 aldehyde to the corresponding carboxylic acid. As mentioned previously, both PQQ-ADHs have  
394 been biochemically shown to act on aldehydes and could catalyze the reaction, but the lack of a  
395 strong fitness phenotype for both ethanol and n-butanol suggest they are not the only enzymes  
396 capable of catalyzing this reaction. The genomically proximal aldehyde dehydrogenase *pedI*  
397 (PP\_2680) showed minor fitness defects when grown on ethanol and several other alcohols  
398 (**Figure 5, Figure 6A**), but showed no fitness defects when libraries were grown on butanol or  
399 pentanol. Another aldehyde dehydrogenase, *aldB-I* (PP\_0545), showed virtually no fitness  
400 defects when grown on any of the short chain n-alcohols tested here (**Figure 6A**). The lack of  
401 any one obvious enzyme with a distinct fitness defect supports the notion that multiple enzymes  
402 are present and able to catalyze the oxidation of these aldehydes.

403         While the metabolism of alcohols and their corresponding fatty acids are similar, their  
404 fitness patterns showed distinct differences. When grown on acetate, mutants in *gacS* or *gacA*  
405 (PP\_1650 and PP\_4099 - a two-component (TCS) system), *sigS* (PP\_1623 - the stationary phase  
406 sigma factor sigma S), and *ptsH* (PP\_0948 - a component of the sugar phosphotransferase system

407 (PTS)) showed large and significant fitness benefits, which were not apparent when grown on  
408 ethanol (**Figure 5B**). The GacS/GacA TCS is widespread across many gram-negative bacteria,  
409 and is believed to exert transcriptional control over a wide variety of functions, sometimes in  
410 concert with a small RNA binding protein (CsrA) that exerts post-transcriptional control (57). In  
411 *Pseudomonads*, the GacA/GacS TCS has been implicated in positively controlling *sigS*  
412 expression in multiple species (58). In *P. putida* specifically, *gacS* mutations in strains  
413 engineered to produce muconic acid have resulted in higher titers (59), but disruption of the gene  
414 was also shown to completely abolish production of medium-chain length  
415 polyhydroxyalkanoates (PHAs) (60). Growth on butyrate also showed that *gacS*, *gacR*, *sigS*, and  
416 another component of the PTS (*ptsP*) had significant fitness benefits if disrupted, which was not  
417 observed when the library was grown on butanol (**Figure 5C**). Interestingly, mutants in *gacA* and  
418 *gacS* seemed to have fitness benefits when grown on either pentanol or valerate (**Figure 5D**).  
419 Further work is necessary to precisely characterize the nature of the benefits that occur when  
420 these genes are disrupted.

421         When grown on ethanol compared to acetate, relatively few genes not involved in the  
422 oxidation of the short chain alcohols were found to be specifically and significantly unfit;  
423 however, specific phenotypes for acetate catabolism were observed (**Figure 5B**). Mutants in  
424 PP\_1635 (a two-component system response regulator), PP\_1695 (variously annotated as a  
425 sodium-solute symporter, sensory box histidine kinase, or response regulator), and *tal* (PP\_2168  
426 - transaldolase) all showed fitness defects on acetate that were not observed when libraries were  
427 grown on ethanol. The high cofitness between PP\_1635 and PP\_1695 observed across all  
428 publicly available fitness data ( $r = 0.88$ ) and share homology to *crbSR* systems of other bacteria  
429 where it is known to regulate acetyl-coA synthetase (61).

430 Much like ethanol and acetate, there were relatively few genes that showed specific  
431 fitness defects when grown on butanol that were not also observed in butyrate. However, the  
432 genes *glgB* (PP\_4058 - a 1,4-alpha-glucan branching enzyme), and the co-transcribed PP\_2354  
433 and PP\_2356 (annotated as a histidine kinase/response regulator (HK/RR), and histidine kinase  
434 respectively) showed specific fitness defects when grown on butyrate relative to butanol.  
435 PaperBLAST analysis of PP\_2356 and PP\_2354 did not reveal any publications that had  
436 explored the function of this system, and thus further work will be needed to better characterize  
437 its regulon (62). Mutants of genes encoding for three TCSs were found to be specifically unfit  
438 when grown on pentanol when compared to valerate. PP\_2683 (a two component HK/RR), and  
439 *pedR1* (PP\_2665 - RR) were both specifically unfit and, as previously described, are involved in  
440 the regulation of the *ped* cluster (**Figure 5D**). The gene *cbrB* (PP\_4696 - sig54-dependent RR)  
441 also showed pentanol-specific defects, and is known to regulate central carbon metabolism and  
442 amino acid uptake in the *Pseudomonads* (63, 64).

#### 443 Short chain diol catabolism

444 Another group of industrially relevant alcohols with potential for biotechnological  
445 production are short chain diols. These compounds have broad utility ranging from plasticizers to  
446 food additives (65). As shown in **Figure 5**, most of the tested short chain diols result in  
447 significant fitness defects in *pedR1*, indicating that some of the genes involved in these  
448 metabolisms are in the PedR1 regulon. However, only 1,5-pentanediol had a strong fitness defect  
449 in *pedF*, indicating that multiple dehydrogenases may act on the shorter chain diols.  
450 Additionally, both 1,2-propanediol and 1,3-butanediol cause slight defects in mutants of the  
451 aldehyde dehydrogenase PP\_0545. Although there is some ambiguity as to which enzymes

452 initially oxidize the diols to their corresponding acids, the remaining steps in 1,2-propanediol,  
453 1,3-butanediol, and 1,5-pentanediol catabolism are much more straightforward.

454       Oxidation of 1,2-propanediol yields lactate, and mutants in the L-lactate permease  
455 PP\_4735 (*lldP*) have a fitness of -4.3 when grown on 1,2-propanediol. Furthermore, under this  
456 condition, mutants of the L- and D- lactate dehydrogenases PP\_4736 (*lldD*) and PP\_4737 (*lldE*)  
457 have fitness defects of -5.0 and -1.5, respectively. Since we provided a rac-1,2-propanediol as a  
458 substrate, this likely explains the fitness defects observed in both dehydrogenases (66, 67). Given  
459 these results, it appears that 1,2-propanediol is assimilated into central metabolism via oxidation  
460 to pyruvate (**Figure S6**).

461       When grown on 1,3-butanediol, two oxidations of 1,3-butanediol result in 3-  
462 hydroxybutyrate, and we observe fitness defects of -2.5 in the D-3-hydroxybutyrate  
463 dehydrogenase PP\_3073 and -1.8 in the neighboring sigma-54 dependent regulator PP\_3075  
464 (68). Dehydrogenation of 3-hydroxybutyrate results in acetoacetate, and we see a fitness defect  
465 of -2.9 and -3.0 for the subunits of the predicted 3-oxoacyl-CoA transferase PP\_3122-3 (*atoAB*).  
466 This enzyme likely transfers a CoA from either succinyl-CoA or acetyl-CoA in order to generate  
467 acetoacetyl-CoA. Regarding transport, mutants in the D-beta-hydroxybutyrate permease  
468 PP\_3074, located in the same operon as the 3-hydroxybutyrate dehydrogenase, have a fitness  
469 defect of -0.9, while mutants in the RarD permease PP\_3776 have a fitness of -1.2.

470       Following oxidation by the aforementioned PQQ-dependent dehydrogenases and  
471 aldehyde dehydrogenases in the periplasm, an oxidized intermediate is likely transported into the  
472 cell for the next steps in the catabolism. This is supported by the observation that mutants of the  
473 predicted dicarboxylate MFS transporter PP\_1400 and its two-component regulator PP\_1401-2  
474 have strong fitness defects on both alpha-ketoglutarate and 1,5-pentanediol. Furthermore, there is

475 a -4.7 fitness defect in mutants of the L-2-hydroxyglutarate oxidase PP\_2910, which catalyzes  
476 the second step in the glutarate hydroxylation pathway of glutarate catabolism. The glutarate  
477 hydroxylase PP\_2909, which catalyzes the first step of this pathway, has a much slighter  
478 negative fitness of -0.6. This is expected, because glutarate can also be catabolized through a  
479 glutaryl-CoA dehydrogenation pathway, so mutants in PP\_2909 can simply divert flux through  
480 the other catabolic route (12). Mutants in PP\_2910 are unable to oxidize L-2-hydroxyglutarate to  
481 alpha-ketoglutarate, and likely accumulate L-2-hydroxyglutarate as a dead-end metabolite.

482 1,4-butanediol catabolism has been previously studied; based on the results of expression  
483 data and adaptive laboratory evolution, Li et al. proposed three potential catabolic pathways for  
484 1,4-butanediol, including a beta-oxidation pathway (**Figure 8**) (30). Their evolved strains had  
485 mutations in the LysR activator PP\_2046 that resulted in overexpression of the beta-oxidation  
486 operon PP\_2047-51 (30). Interestingly, we found that when grown on 1,4 butanediol, transposon  
487 mutants of the acyl-CoA dehydrogenase PP\_2048 had significant fitness benefits and no CoA-  
488 ligase mutants showed significant fitness defects. However, a fitness defect of -1.0 in PP\_0356  
489 (malate synthase) mutants suggests that there may be flux through the beta-oxidation pathway to  
490 glycolic acid and acetyl-CoA. A possible explanation for the positive fitness of PP\_2048 mutants  
491 is that the beta-oxidation pathway is suboptimal in the wild type, and it may be beneficial to  
492 divert flux through the other pathway(s). This same reasoning could also explain the absence of  
493 CoA-ligases with fitness defects; however, this also could be due to the presence of multiple  
494 CoA-ligases capable of catalyzing that step. Mutants of the 3-hydroxyacyl-CoA dehydrogenase  
495 PP\_2047, a *fadB* homolog which likely catalyzes the hydration and dehydrogenation steps to  
496 produce 3-oxo-4-hydroxybutyryl-CoA, had a strong fitness defect. When PP\_2047 is non-  
497 functional, 4-hydroxycrotonyl-CoA likely accumulates as a deadend metabolite resulting in

498 decreased fitness. Li et al. also showed that deletion mutants of PP\_2046 are unable to grow on  
499 1,4-butanediol (30). Our data suggests that this is because PP\_2049 appears to be the main  
500 alcohol dehydrogenase acting on either 1,4-butanediol or 4-hydroxybutyrate, and is in the operon  
501 under the control of PP\_2046. Although our fitness data suggests that both the oxidation to  
502 succinate and beta-oxidation pathways occur, further work is necessary to determine if the  
503 pathway to succinyl-CoA is involved in the catabolism.

#### 504 Branched chain alcohol metabolism

505         Due to their superior biofuel properties, branched chain alcohols have been targets for  
506 metabolic engineering as potential alternatives to ethanol and butanol (69). Our fitness data  
507 suggest that *pedE* and/or *pedH* oxidize 2-methyl-1-butanol to 2-methylbutyrate, which then  
508 undergoes one round of beta-oxidation to produce acetyl-CoA and propionyl-CoA (**Figure S7**).  
509 Most of the genes involved in 2-methylbutyrate beta-oxidation are located in the operon  
510 PP\_2213-PP\_2217. With mutants having a fitness defect of -3.2, PP\_2213 appears to be the main  
511 acyl-CoA ligase acting on 2-methylbutyrate. Mutants in two predicted acyl-CoA  
512 dehydrogenases, PP\_2216 and PP\_0358, show fitness defects of -1.1 and -2.6, respectively. The  
513 enoyl-CoA hydratase PP\_2217 has a fitness defect of -5.7 and the 3-hydroxyacyl-CoA  
514 dehydrogenase PP\_2214 has a fitness defect of -5.6. Finally, the acetyl-CoA acetyltransferase  
515 appears to be PP\_2215, with mutants having a fitness defect of -4.8. We also observed fitness  
516 defects of -1.8 and -1.6 in mutants of the ABC transporters, PP\_5538 and PP\_2667, respectively.  
517 Since 2-methylbutyrate is a known intermediate in the catabolism of isoleucine, we found that  
518 the genetic data presented here closely mirror the previous biochemical characterization of this  
519 system (70, 71).

520 *P. putida* can readily grow on isopentanol and isoprenol but not prenol (**Figure 9A**). All  
521 three of these alcohols have been produced in high titers in *Escherichia coli* and other bacteria  
522 because of their potential to be suitable replacements for gasoline (72, 73). RB-TnSeq data for  
523 isopentanol and isoprenol showed severe fitness defects in genes of the leucine catabolic  
524 pathway (**Figure 10**). This is unsurprising, as isopentanol can be generated from the leucine  
525 biosynthetic pathway (74). Deletion of the PP\_4064-PP\_4067 operon, which contains the genes  
526 that code for the conversion of isovaleryl-CoA to 3-hydroxy-3-methylglutaryl-CoA, abolished  
527 growth on both isopentanol and isoprenol (**Figure S8**). Deletion of PP\_3122 (acetoacetyl CoA-  
528 transferase subunit A) also abolished growth on isopentanol, and greatly reduced growth on  
529 isoprenol (**Figure S8**). Taken together, these results validate that both of these alcohols are  
530 degraded via the leucine catabolic pathway. Transposon insertion mutants in *pedF* showed strong  
531 fitness defects on both isopentanol and isoprenol, suggesting that *pedH* (PP\_2679) and *pedE*  
532 catalyze (PP\_2674) the oxidation of the alcohols. Deletion mutants in *pedH* showed only a minor  
533 delay in growth compared to wild-type when grown on either isopentanol or isoprenol, while  
534 mutants in *pedE* showed a more substantial growth defect on both alcohols (**Figure 9A**).  
535 Deletion of *pedF* (PP\_2675) prevented growth on both isopentanol and nearly abolished growth  
536 on isoprenol when provided as a sole carbon source in minimal media (**Figure 9A**). When wild-  
537 type *P. putida* was grown in minimal media with 10 mM glucose and 4 mM of either  
538 isopentanol, prenol, or isoprenol, each alcohol was shown to be readily degraded with concurrent  
539 observation of increasing levels of the resultant acid (**Figure 9B**). Though *P. putida* was unable  
540 to utilize prenol as a sole carbon source, it was still able to readily oxidize prenol to 3-methyl-2-  
541 butenoic acid, suggesting there is no CoA-ligase present in the cell able to activate this substrate  
542 and channel it towards leucine catabolism (**Figure 10**). When wild-type *P. putida* was grown in



543 LB medium supplemented with 4 mM of each alcohol individually, all alcohols were completely  
544 degraded by 24 hours post-inoculation (**Figure 9C**). In *pedF* deletion mutants grown under the  
545 same conditions, the rate at which the alcohols were degraded was significantly slowed; however  
546 after 48 hours ~50% of isopentanol, ~75% of isoprenol, and 100% of prenol were degraded  
547 (**Figure 9C**). Uninoculated controls showed that no alcohol was lost at greater than 5% on  
548 account of evaporation (data not shown). Future efforts to produce any of these alcohols in *P.*  
549 *putida* will be heavily impacted by this degradation, and greater effort will need to be made to  
550 identify other enzymes involved in the oxidation of these alcohols or other metabolic pathways  
551 that consume them.

552         One mystery that remains is how isoprenol enters into leucine catabolism. GC-MS  
553 analysis confirmed oxidation of the alcohol to 3-methyl-3-butenic acid, but it is unclear what  
554 the next step entails. Fitness data suggests that either PP\_4063 or PP\_4549 may attach the CoA  
555 to isovalerate, but neither of these genes have strong phenotypes when mutant libraries are  
556 grown on isoprenol (**Figure 10**). That PP\_4064 (isovaleryl-CoA dehydrogenase) shows strong  
557 negative fitness values when libraries are grown on isoprenol implies that its degradation goes  
558 through an isovaleryl-CoA intermediate, however this fitness defect may be the result of polar  
559 effects that disrupt the downstream steps (**Figure 10**). One possibility is that 3-methyl-3-  
560 butenoic acid is reduced to isovalerate in the cell; however, this seems unlikely since no  
561 isovalerate was observed via GC-MS when *P. putida* was fed isoprenol and glucose. Two other  
562 possible routes could result from the activation of 3-methyl-3-butenic acid by an undetermined  
563 CoA-ligase. If this CoA-ligase exists, it is interesting that it would have activity on 3-methyl-3-  
564 butenoic acid but not 3-methyl-2-butenic acid, which accumulates when *P. putida* is grown in the  
565 presence of prenol. Once formed, the 3-methyl-3-butenyl-CoA could be directed into leucine

566 catabolism via either an isomerization to 3-methylcrotonyl-CoA or a reduction to isovaleryl-  
567 CoA. Future work that leverages metabolomics to identify compounds that accumulate in leucine  
568 catabolic mutants may reveal the missing steps and help narrow the search for their enzymes.

#### 569 Future Directions

570           The large set of global fitness data generated in this study provide an extensive and  
571 global overview on the putative pathways of alcohol and fatty acid degradation in *P. putida*.  
572 Overall, our fitness data agree with previously published biochemical data that explored enzymes  
573 in both fatty acid and alcohol metabolism. However, there are still many questions that our data  
574 leave unanswered. Further investigation will be required to untangle and elucidate which specific  
575 enzymes are biologically relevant in the beta-oxidation of short chain fatty acids. It is likely that  
576 biochemical characterization of individual enzymes will be required to determine which of the  
577 *fad* homologs catalyze these reactions. Another intriguing question is the function of PP\_0765  
578 and PP\_0766. Biochemical interrogation and mutational analysis of the DUF1302 and DUF1329  
579 family proteins are needed to determine whether these proteins indeed function as an esterase or,  
580 as previously predicted, play some other role in outer membrane biogenesis (41). Additional  
581 work is also warranted to ascertain which of the proposed 1,4-butanediol catabolic routes the  
582 wild-type organism actually uses and determine whether the beta-oxidation pathway is indeed  
583 less preferable than the pathway to succinate.

584           To our knowledge, our finding that *P. putida* can consume both isopentanol and isoprenol  
585 are the first observations of this metabolism. If metabolic engineers wish to produce these  
586 chemicals in *P. putida*, these pathways will need to be removed. Critically, researchers will need  
587 to identify other enzymes that result in the oxidation of these alcohols or other routes of  
588 degradation within *P. putida*. How *P. putida* is able to utilize isoprenol, but not prenol, as a sole

589 carbon source is metabolically intriguing. One of our proposed pathways of isoprenol catabolism  
590 requires the existence of a CoA-ligase that shows surprising specificity towards 3-methyl-3-  
591 butenoic acid with little to no activity on 3-methyl-2-butenic acid. More work should be done to  
592 leverage other omics-levels techniques to try to identify this hypothetical enzyme and  
593 biochemically verify its activity. Finally, this data set as a whole will likely strengthen the  
594 assumptions made by genome-scale metabolic models. Previous models of *P. putida* metabolism  
595 have incorporated RB-TnSeq data to improve their predictions (17). This work nearly doubles  
596 the number of available RB-TnSeq datasets in *P. putida* that are publicly available and will likely  
597 contribute greatly to further model refinement. Ultimately, large strides in our understanding of  
598 *P. putida* metabolism leveraging functional genomic approaches will provide the foundation for  
599 improved metabolic engineering efforts in the future.

600

## 601 **Methods**

### 602 Media, chemicals, and culture conditions

603 General *E. coli* cultures were grown in lysogeny broth (LB) Miller medium (BD  
604 Biosciences, USA) at 37 °C while *P. putida* was grown at 30 °C. When indicated, *P. putida* and  
605 *E. coli* were grown on modified MOPS minimal medium, which is comprised of 32.5 μM CaCl<sub>2</sub>,  
606 0.29 mM K<sub>2</sub>SO<sub>4</sub>, 1.32 mM K<sub>2</sub>HPO<sub>4</sub>, 8 μM FeCl<sub>2</sub>, 40 mM MOPS, 4 mM tricine, 0.01 mM  
607 FeSO<sub>4</sub>, 9.52 mM NH<sub>4</sub>Cl, 0.52 mM MgCl<sub>2</sub>, 50 mM NaCl, 0.03 μM (NH<sub>4</sub>)<sub>6</sub>Mo<sub>7</sub>O<sub>24</sub>, 4 μM H<sub>3</sub>BO<sub>3</sub>,  
608 0.3 μM CoCl<sub>2</sub>, 0.1 μM CuSO<sub>4</sub>, 0.8 μM MnCl<sub>2</sub>, and 0.1 μM ZnSO<sub>4</sub> (75). Cultures were  
609 supplemented with kanamycin (50 mg/L, Sigma Aldrich, USA), gentamicin (30 mg/L, Fisher  
610 Scientific, USA), or carbenicillin (100mg/L, Sigma Aldrich, USA), when indicated. All other

611 compounds were purchased through Sigma Aldrich (Sigma Aldrich, USA). 3-methyl-3-butenic  
612 acid was not available commercially and required synthesis which is described below.

### 613 Strains and plasmids

614 All bacterial strains used in this study can be found in Table 1 and plasmids used in this  
615 work are listed in Table 2. All strains and plasmids created in this work are available through the  
616 public instance of the JBEI registry. ([public-registry.jbei.org/folders/456](http://public-registry.jbei.org/folders/456)). All plasmids were  
617 designed using Device Editor and Vector Editor software, while all primers used for the  
618 construction of plasmids were designed using j5 software (76–78). Plasmids were assembled via  
619 Gibson Assembly using standard protocols (79), or Golden Gate Assembly using standard  
620 protocols (80). Plasmids were routinely isolated using the Qiaprep Spin Miniprep kit (Qiagen,  
621 USA), and all primers were purchased from Integrated DNA Technologies (IDT, Coralville, IA).  
622 Construction of *P. putida* deletion mutants was performed as described previously (18).

### 623 Plate-based growth assays

624 Growth studies of bacterial strains were conducted using microplate reader kinetic assays  
625 as described previously (82). Overnight cultures were inoculated into 10 mL of LB medium from  
626 single colonies, and grown at 30 °C. These cultures were then washed twice with MOPS minimal  
627 media without any added carbon and diluted 1:100 into 500 µL of MOPS medium with 10 mM  
628 of a carbon source in 48-well plates (Falcon, 353072). Plates were sealed with a gas-permeable  
629 microplate adhesive film (VWR, USA), and then optical density and fluorescence were  
630 monitored for 48 hours in an Biotek Synergy 4 plate reader (BioTek, USA) at 30 °C with fast  
631 continuous shaking. Optical density was measured at 600 nm.

### 632 RB-TnSeq

633 RB-TnSeq experiments utilized *P. putida* library JBEI-1 which has been described  
634 previously with slight modification(18). Libraries of JBEI-1 were thawed on ice, diluted into 25  
635 mL of LB medium with kanamycin and then grown to an OD<sub>600</sub> of 0.5 at 30 °C at which point  
636 three 1-mL aliquots were removed, pelleted, and stored at -80 °C. Libraries were then washed  
637 once in MOPS minimal medium with no carbon source, and then diluted 1:50 in MOPS minimal  
638 medium with 10 mM of each carbon source tested. Cells were grown in 10 mL of medium in test  
639 tubes at 30 °C shaking at 200 rpm. One 500-μL aliquot was pelleted, and stored at -80 °C until  
640 BarSeq analysis, which was performed as previously described (19, 40). The fitness of a strain is  
641 defined here as the normalized log<sub>2</sub> ratio of barcode reads in the experimental sample to barcode  
642 reads in the time zero sample. The fitness of a gene is defined here as the weighted average of  
643 the strain fitness for insertions in the central 10% to 90% of the gene. The gene fitness values are  
644 normalized such that the typical gene has a fitness of zero. The primary statistic *t* value  
645 represents the form of fitness divided by the estimated variance across different mutants of the  
646 same gene. Statistic *t* values of  $>|4|$  were considered significant. A more detailed explanation of  
647 calculating fitness scores can be found in Wetmore et al. (40). All experiments described here  
648 passed testing using the quality metrics described previously unless noted otherwise (40). All  
649 experiments were conducted in biological duplicate, and all fitness data are publically available  
650 at <http://fit.genomics.lbl.gov>.

#### 651 GC-MS and GC-FID Analysis of Branched Alcohol Consumption

652 To examine the oxidation of isopentanol, prenol, and isoprenol to their corresponding  
653 acids 10mL of MOPS minimal medium supplemented with 10 mM glucose and 4mM of one of  
654 each alcohol added were inoculated with a 1:100 dilution of overnight *P. putida* culture and  
655 incubated at 30 °C with 200 rpm shaking. At 0, 12, 24, and 48-hours post-inoculation 200 μL of

656 media were sampled and stored at - 80 °C. Alcohols and fatty acids were extracted by acidifying  
657 media with 10 µL of 10N HCl, followed by addition of an 200 µL of ethyl-acetate. To detect  
658 alcohols and their corresponding carboxylic acids via GC-MS an Agilent 6890 system equipped  
659 with a DB-5ms column (30- m×0.25 mm×0.25 µm) and an Agilent 5973 MS were used. Helium  
660 (constant flow 1 mL/min) was used as the carrier gas. The temperature of the injector was 250 °C  
661 and the following temperature program was applied: 40 °C for 2 min, increase of 10 °C/min to  
662 100 °C then increase of 35 °C/min to 300 °C , temperature was then held at 300 °C for 1 min.  
663 Authentic standards were used to quantify analytes. Determination of isopentanol, prenol, and  
664 isoprenol consumption was conducted in 10mL LB medium with 4mM of either alcohol added.  
665 Cultures were inoculated with a 1:100 dilution of overnight *P. putida* culture and incubated at  
666 30 °C with 200 rpm shaking. At 0, 24, and 48 hours post-inoculation 200 µL of media were  
667 sampled and stored at - 80 °C. The remaining concentration of each alcohol was determined by  
668 GC-FID as previously described (83).

#### 669 Synthesis of 3-Methyl-3-Butenoic Acid

670 To a 25-mL round bottom flask was added chromium(VI) oxide (0.69 g, 6.9 mmol) and distilled  
671 water (1 mL). The reaction mixture was then cooled to 0 °C before concentrated sulfuric acid (0.6  
672 mL, 10.5 mmol) was added dropwise, thus forming Jones reagent. The solution of Jones reagent  
673 was then diluted to a total volume of 5 mL with distilled water. To a stirred solution of 3-methyl-  
674 3-buten-1-ol (0.59 g, 6.9 mmol) in acetone (7 ml) was added dropwise the Jones reagent at 0 °C.  
675 After being stirred for 8 h at room temperature, the mixture was quenched with ethanol. The  
676 mixture was then diluted with water, and acetone was evaporated *in vacuo*. The residue was  
677 extracted with DCM, and organic layers were combined and washed three times with saturated *aq.*  
678 NaHCO<sub>3</sub> solution. The aqueous phase was acidified with a 2 M *aq.* HCl solution to pH 2-3, which

679 was then extracted again with DCM. The extract was successively washed with water and brine,  
680 dried over MgSO<sub>4</sub>, and concentrated *in vacuo*. The residue was distilled (90 °C, 100 mTorr) to  
681 yield 3-methyl-3-butenic acid as a clear oil. <sup>1</sup>H NMR (300 MHz, Chloroform-*d*) δ 4.92 (d, *J* =  
682 19.1 Hz, 2H), 3.08 (s, 2H), 1.84 (s, 3H) (**Figure 11**).

683

#### 684 Bioinformatic Analyses

685 PaperBLAST was routinely used to search for literature on proteins of interest and related  
686 homologs (62). All statistical analyses were carried out using either the Python Scipy or Numpy  
687 libraries (84, 85). For the phylogenetic reconstructions, the best amino acid substitution model  
688 was selected using ModelFinder as implemented on IQ-tree (86) phylogenetic trees were  
689 constructed using IQ-tree, nodes were supported with 10,000 bootstrap replicates. The final tree  
690 figures were edited using FigTree v1.4.3 (<http://tree.bio.ed.ac.uk/software/figtree/>). Orthologous  
691 syntenic regions were identified with CORASON-BGC (87) and manually colored and  
692 annotated.

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714

## 715 **Contributions**

716 Conceptualization, M.G.T., M.R.I., A.N.P.; Methodology, M.G.T., M.R.I., A.N.P., J.M.B,  
717 P.C.M., A.M.D.; Investigation, M.G.T., M.R.I., A.N.P, M.S., W.A.S., C.B.E., P.C.M., J.M.B.,  
718 Y.L., R.W.H., C.A.A, R.N.K, P.L.; Writing – Original Draft, M.G.T., M.R.I., A.N.P.; Writing –  
719 Review and Editing, All authors.; Resources and supervision, L.M.B., A.M., A.M.D., P.M.S,  
720 J.D.K.

721 M.G.T., M.R.I., and A.N.P. contributed equally to this work. Author order was determined by  
722 the outcome of a MarioKart 64 tournament.



723 **Competing Interests**

724 J.D.K. has financial interests in Amyris, Lygos, Demetrix, Napigen, Maple Bio, and Apertor  
725 Labs. C.B.E has a financial interest in Perlumi Chemicals.

726

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1006 **Table 1: Strains used in this study.**

| Strain                            | Description                                                                | Reference  |
|-----------------------------------|----------------------------------------------------------------------------|------------|
| <i>E. coli</i> XL1 Blue           |                                                                            | Agilent    |
| <i>P. putida</i> KT2440           | Wild-Type                                                                  | ATCC 47054 |
| <i>P. putida</i> ΔPP_2674         | Strain with complete internal in-frame deletion of PP_2674                 | This study |
| <i>P. putida</i> ΔPP_2675         | Strain with complete internal in-frame deletion of PP_2675                 | This study |
| <i>P. putida</i> ΔPP_2679         | Strain with complete internal in-frame deletion of PP_2679                 | This study |
| <i>P. putida</i> ΔPP_3839         | Strain with complete internal in-frame deletion of PP_3839                 | This study |
| <i>P. putida</i> ΔPP_2675ΔPP_3839 | A double knockout of PP_2675 and PP_3839                                   | This study |
| <i>P. putida</i> ΔPP_4064-PP_4067 | Strain with complete internal in-frame deletion of the PP_4064-4067 operon | This study |
| <i>P. putida</i> ΔPP_3122         | Strain with complete internal in-frame deletion of PP_3122                 | This study |

1007

1008 **Table 2: Plasmids used in this study.**

| Plasmid                | Description                                                            | Reference          |
|------------------------|------------------------------------------------------------------------|--------------------|
| pMQ30                  | Suicide vector for allelic replace Gm <sup>r</sup> , SacB              | <a href="#">81</a> |
| pMQ30 ΔPP_2674         | pMQ30 derivative harboring 1kb flanking regions of PP_2674             | This study         |
| pMQ30 ΔPP_2675         | pMQ30 derivative harboring 1kb flanking regions of PP_2675             | This study         |
| pMQ30 ΔPP_2679         | pMQ30 derivative harboring 1kb flanking regions of PP_2679             | This study         |
| pMQ30 ΔPP_3839         | pMQ30 derivative harboring 1kb flanking regions of PP_3839             | This study         |
| pMQ30 ΔPP_4064-PP_4067 | pMQ30 derivative harboring 1kb flanking regions of PP_4064 and PP_4067 | This study         |
| pMQ30 ΔPP_3122         | pMQ30 derivative harboring 1kb flanking regions of PP_3122             | This study         |

1009

1010 **Figure 1: cladogram correlation matrix of genome-wide fitness data of *P. putida* grown on fatty acids.** The

1011 matrix shows pairwise comparisons of Pearson correlations of fitness data from *P. putida* KT2440 RB-TnSeq

1012 libraries grown on fatty acids as well as glucose. The legend in top left shows Pearson correlation between two  
1013 conditions with blue showing  $r = 1$ , and red showing  $r = 0$ . The conditions were tested in duplicate and the data  
1014 from each are numbered (1 & 2).

1015  
1016 **Figure 2: Overview of fatty acid catabolic pathways of *P. putida* KT2440.** The above diagram shows the  
1017 catabolic steps of fatty ester and saturated/unsaturated fatty acid catabolism in *P. putida* KT2440, in addition to their  
1018 connections to the glyoxylate shunt and the methylcitrate cycle. The heatmaps below show fitness scores when  
1019 grown on fatty acids or glucose for the specific genes proposed to catalyze individual chemical reactions. Colors  
1020 represent fitness scores, with blue representing positive fitness and red representing negative fitness.

1021  
1022 **Figure 3: Putative pathways for short chain fatty acid catabolism in *P. putida* KT2440.** A) Individual enzymatic  
1023 steps that potentially catalyze the steps of beta-oxidation for short chain fatty acids, fitness scores are listed to the  
1024 right of each enzyme when grown on either butyrate, valerate, or hexanoate. B) The operonic structure of *btkB* and  
1025 *hdb* flanked by an AraC-family (PP\_3753) and TetR-family (PP\_3756). The heatmap shows fitness scores of the  
1026 genes when grown on butyrate, butanol, or levulinic acid.

1027  
1028 **Figure 4: Global analysis of alcohol metabolism in *P. putida*.** A) Pairwise comparisons of Pearson correlations of  
1029 fitness data from *P. putida* KT2440 RB-TnSeq libraries grown on alcohols as well as glucose grouped by overall  
1030 similarity. Colors bar at top left shows the Pearson coefficient with 1 indicating greater similarity and 0 indicating  
1031 greater dissimilarity. B) Heatmap shows the fitness scores of all alcohol dehydrogenases annotated on the BioCyc  
1032 database as well as the cytochrome C PP\_2675 when grown on various alcohols and glucose. C) Operonic diagram  
1033 of the *pqq* cluster in *P. putida* and the corresponding biosynthetic pathway for the PQQ cofactor and D) How PQQ  
1034 cofactors are regenerated by cytochrome C. Heatmap shows fitness scores for individual *pqq* cluster genes when  
1035 grown on alcohols and glucose.

1036  
1037 **Figure 5: Essentiality and regulation of the *ped* cluster.** (Top) Heatmap depicting the fitness scores for genes in  
1038 the *ped* cluster (PP\_2662 to PP\_2683) during growth on various short chain alcohols. (Bottom) Genomic context for

1039 the *ped* cluster in *P. putida* KT2440. Arrows depict transcriptionally upregulated genes of *pedR1* and *pedR2*. Blunt  
1040 arrows point to genes predicted to be transcriptionally repressed in the condition tested.

1041 **Figure 6: Analysis of short chain alcohol metabolism in *P. putida*:** A) Putative genes involved in the initial  
1042 oxidation steps of short chain alcohol assimilation in *P. putida*. PP\_2675 (PedF) is involved in the regeneration of  
1043 the PQQ cofactor predicted to be necessary for these oxidation reactions of PP\_2764 (PedE) and PP\_2769 (PedH).  
1044 Average fitness scores for two biological reps are shown next to each gene for ethanol (black), butanol (green), and  
1045 pentanol (blue). Scatter plots show global fitness scores for ethanol versus acetate (B), butanol versus butyrate (C),  
1046 and pentanol versus valerate (D).

1047  
1048 **Figure 7: Validation of alcohol dehydrogenases involved in short chain alcohol metabolism** Growth curves of  
1049 wild type (blue),  $\Delta$ PP\_2675 (orange),  $\Delta$ PP\_3839 (green), and  $\Delta$ PP\_2675 $\Delta$ PP\_3839 (red) strains of *P. putida*  
1050 KT2440 on 10 mM ethanol (A), 10 mM n-butanol (B), and 10 mM n-pentanol (C). Shaded area represents 95%  
1051 confidence intervals (cI), n=3.

1052  
1053 **Figure 8: Putative routes of 1,4-butanediol catabolism in *P. putida*.** Putative genes involved in catabolism of 1,4-  
1054 butanediol in *P. putida*. Average fitness scores for two biological reps are shown next to each gene. The three CoA-  
1055 ligases shown were proposed by Li et al.; there were no CoA-ligases that showed significant fitness defects on 1,4-  
1056 butanediol. \*PP\_2675 (PedF) is involved in the regeneration of the PQQ cofactor predicted to be necessary for these  
1057 oxidation reactions of PP\_2764 (PedE) and PP\_2769 (PedH).

1058  
1059 **Figure 9: Isopentanol, Prenol, and Isoprenol consumption by *P. putida*.** A) Growth curves of wild type (blue),  
1060 and  $\Delta$ PP\_2674 (orange),  $\Delta$ PP\_2675 (green), and  $\Delta$ PP\_2679 (red) strains of *P. putida* on isopentanol (left), prenol  
1061 (middle), and isoprenol (right). Structure of alcohols are shown above graphs. Shaded area represents 95%  
1062 confidence intervals (cI), n=3. B) Concentrations of alcohols consumed and their corresponding carboxylic acids  
1063 produced over time by wild type. Left panel shows isopentanol and isovalerate, middle panel shows prenol and 3-  
1064 methyl-2-butenic acid, and the right panel shows isoprenol and 3-methyl-3-butenic acid. Structures of  
1065 corresponding carboxylic acids derived from alcohol are shown in graphs. Error bars represent 95% cI, n=3. C)



1066 Consumption of isopentanol (left), prenol (middle), and isoprenol (right) by wild type and  $\Delta$ PP\_2675 *P. putida* over  
1067 time. Error bars represent 95% cI, n=3.

1068

1069 **Figure 10: Putative routes of isopentanol and isoprenol catabolism in *P. putida*.** Diagram shows the proposed  
1070 pathways for the catabolism of isopentanol and isoprenol. Average fitness scores of two biological replicates for  
1071 individual genes can be found next to each gene. Fitness values for isopentanol are shown in blue, while fitness  
1072 values for isoprenol and shown in green. Potential reactions that would bring isoprenol into leucine catabolism are  
1073 marked with a question mark.

1074

1075 **Figure 11: NMR validation of 3-methyl-3-butenic acid.**

1076