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Article

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1 **Prediction of the formation of biogenic non-extractable residues during degradation of**
2 **environmental chemicals from biomass yields**

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11

12

13 **Abstract**

14 Degradation tests with radio or stable isotope labeled compounds enable the detection of the formation of
15 non-extractable residues (NER). In PBT and vPvB assessment, remobilisable NER are considered as a
16 potential risk while biogenic NER from incorporation of labeled carbon into microbial biomass are treated
17 as degradation products. Relationships between yield, released CO₂ (as indicator of microbial activity and
18 mineralization) and microbial growth can be used to estimate the formation of biogenic NER. We provide a
19 new approach for calculation of potential substrate transformation to microbial biomass (theoretical yield)
20 based on Gibbs free energy and microbially available electrons. We compare estimated theoretical yields
21 of biotechnological substrates and of chemicals of environmental concern with experimentally determined
22 yields for validation of the presented approach. A five-compartment dynamic model is applied to simulate
23 experiments of ¹³C-labeled 2,4-D and ibuprofen turnover. The results show that bioNER increases with
24 time, and that most bioNER originates from microbial proteins. Simulations with pre-calculated input data
25 demonstrate that pre-calculation of yields reduces the number of fit parameters considerably, increases
26 confidence in fitted kinetic data and reduces the uncertainty of the simulation results.

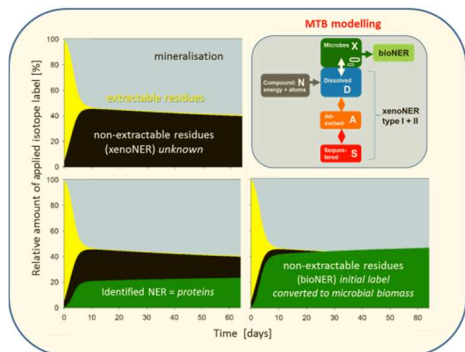
27

28 **Key words:** bound residues, modeling, Gibbs free energy, pesticides, carbon conversion, carbon
29 turnover, microbial biomass, Nernst, NER assessment, OECD tests.

30

31 TOC Art

32



33

34

35 Introduction

36 Degradation is a key parameter in risk assessment and registration of industrial chemicals, veterinary
37 medicinal products and pesticides.¹⁻⁵ Microbial degradability tests are often performed with radio-labeled
38 tracer compounds. Guidelines have been developed for fate assessment in water, sediments and soil,
39 e.g., OECD 304, 307, 308 and 309.⁶⁻⁹ For the interpretation of results, concepts for modeling the turnover
40 kinetics have been developed.^{10,11} Unfortunately, there is still no robust and reliable way to predict the fate
41 of organic molecules in environmental matrices in terms of biotic transformation, mineralization,
42 conversion to microbial biomass and the formation of so-called non-extractable residues (NER).¹²

43 Chemicals may persist in the environment due to several reasons. Relatively well studied is the
44 persistence due to limited bioavailability. Chemicals being strongly adsorbed or sequestered in soil and
45 sediments are often not available for biodegradation.¹²⁻¹⁵ Examples are the five- or six-ring polycyclic
46 aromatic hydrocarbons. Chemicals newly introduced to the biosphere may persist due to the absence of
47 enzymes capable of transforming such compounds. However, after some time for adaptation microbes
48 can “learn” to degrade recalcitrant compounds.¹⁶ A third reason for persistence is that chemicals are poor
49 growth substrates because they do not provide energy, carbon or nutrients to microbes under the specific
50 environmental conditions. For example, alkanes have persisted over millions of years in reservoirs where
51 no suitable electron acceptor (oxygen) was available. Under aerobic conditions, alkanes are excellent
52 substrates with higher microbial biomass yields than glucose.^{17,18} For chlorinated solvents, e.g.
53 trichloroethylene (TCE), the opposite was observed: the chemical provides no energy to microbes under
54 aerobic conditions and is therefore quite persistent, while it can be reductively dehalogenated as electron
55 acceptor in anaerobic groundwater.¹⁹ Another reason for persistence of chemicals can be toxic or
56 inhibitory effects on the microorganisms.^{20,21}

57 In the PBT assessment of industrial chemicals and of veterinary pharmaceuticals, NER are differentiated
58 into remobilisable and irreversibly bound fractions. The irreversibly bound fraction is assessed as a
59 potential removal pathway, while the remobilisable fraction is considered a potential risk for the
60 environment.⁵ Remobilisable NER are sequestered compounds (type I NER) and covalently bound parent
61 compounds or metabolites (NER type II), which may be slowly released. The third fraction is labeled
62 carbon, or other essential elements like nitrogen, transferred to living or dead biomass and eventually

63 fixed in soil organic matter (SOM) derived from decaying microbial biomass (type III NER). These biogenic
64 NER (bioNER) do not constitute any risk.¹² There is thus a need to distinguish harmless, irreversibly
65 bound bioNER from potentially toxic and remobilisable NER (i.e., type I and type II NER) in the risk
66 assessment of chemicals.

67 Determining the microbial biomass yield derived from degradation of a chemical sheds light into the 'black
68 box' of NER. The microbial yield is defined as mass of microbes that can grow on a given amount of
69 substrate (unit g microbial biomass dry weight per g substrate, g g⁻¹).¹⁸ However, most studies with
70 labeled carbon compounds typically express results as g C per g C, and we report these values with their
71 original unit. The unit conversion is shown in the SI. The yield multiplied by the enzymatic substrate
72 removal determines the growth rate of a microbe. High yield can therefore be an indication for the
73 biodegradability of a substrate. Several methods to estimate theoretical microbial yields of a substrate
74 from its energy of formation (Gibbs free energy) have been developed.²²⁻²⁶

75 The yield can be used to predict the likely range of bioNER formed during degradation of environmental
76 chemicals. We i) provide a relationship between formation of bioNER, CO₂ release and yield; ii) present a
77 new and pathway-independent method to estimate yields from thermodynamics combined with an
78 approach to account for the electrons usable by degrading microbes; iii) confirm the yield estimates by
79 comparison to results derived with existing methods²³ and to measured yields of easily degradable
80 carbohydrates, pesticides and other chemicals of environmental concern. iv) Finally, we use the estimated
81 yields as input to the simulation of 2,4-D and ibuprofen biodegradation under formation of microbial
82 biomass, study the performance of the simulation, and compare pre-calculated kinetic parameters with
83 data derived by pure model fit. Data were provided from experimental degradation studies with multi-
84 labeled compounds (¹⁴C, ¹³C) in soil.²⁷⁻²⁹

85

86 **Methods**

87 *Theoretical background*

88 Enzymatic reactions are typically described by the *Michaelis-Menten* equation.^{30,31}

89
$$\frac{dm}{dt} = v_{\max} \times \frac{a}{K_M + a} \times X \quad (\text{eq. 1})$$

90 where m is the mass of chemical substrate metabolized (g), X is the bacterial mass (g bacteria), t is time
91 (d), v_{\max} (g substrate g bacteria⁻¹ d⁻¹) is the maximal substrate consumption rate, a is the chemical activity
92 of the substrate (equivalent to the truly dissolved concentration) (g m⁻³);³² K_M (g m⁻³) is the chemical
93 activity at which the substrate consumption rate is half of its maximum (half saturation or *Michaelis-*
94 *Menten* constant).

95 The yield Y (g g⁻¹) connects metabolism and growth:

96
$$\mu_{\max} = v_{\max} \times Y \quad (\text{eq. 2})$$

97 where μ_{\max} is the maximum growth rate (d⁻¹). Microbes use part of the energy gained from the substrate
98 for growth, and part for maintenance purposes. Experimentally observed net yields equal the true yield
99 minus cell decay. Introducing a term for cell decay or maintenance into the *Monod* equation for microbial
100 growth leads to eq. 3.³³

101
$$\frac{dX}{dt} = \frac{\mu_{\max} a}{K_M + a} \times X - b \times X \quad (\text{eq. 3})$$

102 where dX/dt is the change of microbial biomass with time (g microbial biomass dw per day), and b is the
103 decay rate of microbes (death rate, d⁻¹). For the calculation of the fate of chemicals in soils or sediments, a
104 two-compartment-sorption model^{34,35} calculating rapid (adsorption) and slow (sequestration) kinetics was
105 combined with the equations for microbial metabolism and growth (eqs. 1 and 3).^{12,36} The complete model
106 is described in the Supporting Information (SI).

107

108 *Carbon budget and calculation of bioNER*

109 Few experimental studies deliver compound concentrations and biomass formation in a resolution that
110 allows fitting of dynamic models to the data. In degradation studies according to OECD guidelines, only
111 the fractions of NER, CO₂ and metabolites at the end of the experiment are reported.³⁷ Some general

112 rules and patterns can be derived concerning the distribution of the initially applied labeled carbon and the
 113 formation of biomass (here all units are g C).

114 We define S as the total mineralized substrate, S is initial amount of labeled carbon minus non-
 115 metabolized parent compound minus intermediate metabolites and minus $NER^{I,II}$. $NER^{I,II}$ denotes non-
 116 extractable residues due to sequestration (I) and co-valent binding (II).¹² The biomass produced from
 117 mineralization of the substrate is per definition the yield, hence, as long as growth alone is considered, $X =$
 118 $Y S$. The remaining labeled carbon is oxidized to carbon dioxide, thus $CO_2 = (1-Y) S$. Under these
 119 assumptions, the ratio of X to CO_2 is

$$120 \quad \frac{[X]}{[CO_2]} = \frac{Y}{(1-Y)} \quad (\text{eq. 4})$$

121 The labeled carbon fixed in biomass due to substrate mineralization is part of the bioNER. Eq. 4 does not
 122 take decay of biomass into consideration. Earlier long-term studies³⁸⁻⁴⁰ over 224 days showed that
 123 microbial necromass is a significant source of non-living soil organic matter. In these experiments,
 124 approximately 40% of the labeled carbon initially fixed in biomass X (mainly the protein fraction) turned
 125 into SOM (which also is part of bioNER) and 10% remained within living biomass X. It follows that for $t \rightarrow$
 126 ∞ the fraction f (approximately 0.5) of the decaying X turns into bioNER, and $1-f$ forms CO_2 .⁴⁰ The
 127 resulting ratio of bioNER to CO_2 in long-term experiments with decomposition of dead biomass (and
 128 neglecting slow decomposition of SOM to CO_2) is

$$129 \quad \frac{[bioNER]}{[CO_2]} = \frac{f \times Y}{(1-Y) + (1-f) \times Y} \quad (\text{eq. 5})$$

130

131 *Yield estimates*

132 The microbial yield of a substrate can be estimated from thermodynamics or from empirical equations.
 133 Approaches for yield estimation have been presented and tested by a number of researchers.^{22-24, 26,41} The
 134 approaches of McCarty²³ and Xiao and vanBriesen²⁶ require information on the metabolic pathway of the
 135 compound, which is often not given for environmental chemicals' degradation. Therefore, we based our
 136 estimates of yields on a modified approach of Diekert²², which uses the Gibbs free energy of formation

137 and the structural formula. We modified the method by specifying how much of the formation energy can
 138 be used by microbes.

139
 140 **Microbial Turnover to Biomass (MTB) - a pathway-independent thermodynamic yield estimation method**

141 The approach considers that a substrate can be utilized for anabolism and catabolism:

$$142 \quad N = N_{\text{anabolic}} + N_{\text{catabolic}} = 1/Y \quad (\text{eq. 6})$$

143 where N is the nutritional value (g substrate needed per g microorganism formed), the inverse of yield Y :

$$144 \quad \frac{1}{Y} = \frac{1}{Y_{\text{anabolic}}} + \frac{1}{Y_{\text{catabolic}}} \quad (\text{eq. 7})$$

145
 146 Anabolism

147 The yield associated to anabolism is

$$148 \quad Y_{\text{anabolic}} = \frac{n_c \times M_C}{f_C \times M_S} \quad [\text{g microbial biomass dw g}^{-1} \text{ substrate}] \quad (\text{eq. 8})$$

149 where n_c is mol C per mol substrate, M_C and M_S are the molar masses of carbon (index C) and of
 150 substrate (index S), respectively, and f_C is fraction of C in bacterial dry weight (default 0.53 g carbon g⁻¹
 151 microbial biomass dw).²³

152
 153 Catabolism

154 The yield due to catabolic energy gain can be calculated in five steps.

155 Step i) Free energy of the reaction: The free energy of the reaction (change of Gibbs free energy, kJ mol⁻¹)
 156 is the sum of the Gibbs free energy of formation of products minus educts:

$$157 \quad \Delta G_r^{0'} = \sum \Delta G_f^{0'} \text{ products} - \sum \Delta G_f^{0'} \text{ educts} \quad (\text{eq. 9})$$

158 where $G^{0'}$ is the Gibbs free energy (subscript f for formation, r for reaction) at standard-state conditions (1
 159 mol L⁻¹, indicated by superscript 0) and at a pH of 7 (indicated by superscript '). At activities differing from
 160 1 mol L⁻¹, the change of Gibbs free energy of the reaction $\Delta G_r'$ is

$$161 \quad \Delta G_r' = \Delta G_r^{0'} + RT \ln \left(\frac{[\text{products}]}{[\text{educts}]} \right) \quad (\text{eq. 10})$$

162 where R (8.314 J mol⁻¹ K⁻¹) is the universal gas constant and T (K) is the absolute temperature.

163 Step ii) Electron transfer during the reaction (*Nernst* equation): The *Nernst* equation states that the change
 164 of Gibbs free energy ΔG_r is related to the number of electrons n transferred during the reaction, and the
 165 redox potential E (V) of the reaction:

$$166 \quad \Delta G_r = n \times F \times E \quad (\text{eq. 11})$$

167 where F is the Faraday constant. The number of electrons n transferred in the reaction can be calculated
 168 from the change of the oxidation state (OS) of carbon during the reaction,

$$169 \quad n = (\text{OS}_{\text{Product}} - \text{OS}_{\text{Substrate}}) \times n_C \quad (\text{eq. 12})$$

170 where n_C is again the carbon atoms in the substrate (mol C per mol substrate) which is the same as the
 171 moles of CO₂-molecules formed during complete mineralization. The oxidation state of carbon in the
 172 substrate is:

$$173 \quad \text{OS}_{\text{Substrate}} = \frac{-1 \times H + 2 \times O + 3 \times N - 3 \times P + 2 \times S + 1 \times Cl}{n_C} \quad (\text{eq. 13})$$

174 where the letters stand for the number of the respective atoms in the molecule. After complete oxidation to
 175 CO₂, the carbon in the product has the oxidation state 4, hence OS_{Product} = +4.

176 Step iii) Energy available for the microbe: During biological oxidation, the organisms can use only some
 177 types of electron transfers. The free energy of the reaction is thus the upper limit ("maximum") for the
 178 energy that can be provided by the chemical. The maximum energy gained by the organism during
 179 catabolism may be considerably lower than that. As a general rule, when compounds containing hydrogen
 180 atoms connected to carbon atoms are oxidized to CO₂ and H₂O, the electrons transferred in this reaction

181 are available for microbes to gain energy, i.e. 2 electrons per C–H bond. Thus the number of electron
 182 transfers that can at least be used by microorganisms in a redox reaction is $n_{bio} \geq 2 \times H$ (only H bound to C
 183 atoms are counted). Subsequently, the minimum energy available for ATP synthesis by an organism
 184 ($\Delta G'_{bio}$) is:

$$185 \quad \Delta G'_{bio} = \frac{n_{bio}}{n} \Delta G'_r \quad (\text{eq. 14})$$

186 Step iv) ATP production: With an efficiency of 40% of the microbial catabolism,²² the synthesis of 1 mol
 187 ATP from 1 mol ADP requires 80 kJ (the Gibbs free energy of the reaction is -32 kJ mol^{-1}).⁴¹ Thus, the
 188 microbes can generate β mol ATP per mol substrate:

$$189 \quad \beta = \frac{\Delta G'_{bio}}{-80 \text{ kJ/mol}} \quad (\text{eq. 15})$$

190 Step v) Catabolic energy is used for the formation of new cell material: The produced ATP provides the
 191 energy to form new cell material.⁴¹ Y_{ATP} is the microbial biomass dw that can be formed per mol ATP.⁴¹
 192 Diekert²² provided a range from 2 (CO_2) to 12 (glucose) g microbial biomass dw per mol ATP. Hence, we
 193 use the value of 5 g microbial biomass dw mol^{-1} ATP as default for xenobiotic chemicals, but higher values
 194 for compounds similar to glucose (for details, see SI). The yield due to catabolic energy gain can finally be
 195 calculated by

$$196 \quad Y_{catabolic} = \frac{\beta \times Y_{ATP}}{M_S} \quad [\text{g microbial biomass dw g}^{-1} \text{ substrate}] \quad (\text{eq. 16})$$

197

198 The five steps can be summarized in one equation:

$$199 \quad Y_{catabolic} = \frac{\beta \times Y_{ATP}}{M_S} = \frac{n_{bio}}{n} \times \frac{\Delta G'_r}{-80 \text{ kJ/mol}} \times \frac{Y_{ATP}}{M_S} \quad (\text{eq. 17})$$

200

201 The more detailed approaches of McCarty²³ and Xiao and vanBriesen²⁶ estimate β dependent on the
202 biochemical pathway. Knowledge of the pathway is not required in the method presented because all
203 substrate used for catabolic yield is completely oxidized to CO₂.

204

205 *Choice of compounds for yield estimation*

206 The estimation of yields is commonly applied in biotechnology or wastewater treatment. In environmental
207 chemistry, it has been used by Helbling *et al.*⁴² to estimate the yields of two pesticides and by Yuan and
208 vanBriesen⁴³ to estimate the yield of two chelating agents. First, we investigated the performance and the
209 variance of results of the estimation methods with common substrates in biotechnological applications, for
210 which measured yields are widely available. Second, we applied the method to a set of chemicals of
211 environmental concern. The selection of chemicals for this study was based on: i) availability of measured
212 data on bioNER (2,4-D, ibuprofen), ii) availability of biomass yield data (nitrilotriacetic acid [NTA], linuron,
213 carbofuran, toluene), iii) knowledge on specific degradation pathways, electron acceptors or persistence
214 (pentachlorophenol PCP, carbon tetrachloride, trichloroethene, DDT), and iv) availability of Gibbs free
215 energies of formation (Table S2).

216

217 *Brief description of experiments*

218 Nowak *et al.*^{27,28} thoroughly balanced the formation of bioNER in a fate study with the ¹³C-labeled
219 pesticide 2,4-D and the medical drug ibuprofen in soil. The authors also analysed the amount of ¹³C
220 converted to total amino acids (tAA), total fatty and phospholipid fatty acids (PLFA). The tAA increased
221 over time although the PLFA as marker for living biomass declined already after one (2,4-D) to three
222 (ibuprofen) weeks. The details of the turnover experiments are provided in Nowak *et al.*^{27,28}; the results of
223 the experiments are shown in Tables S9 and S10.

224

225 *Description of simulations*

226 2,4-D and ibuprofen experiments were simulated to confirm the relation between yield and bioNER
 227 formation. For a detailed description of model and input data see SI S2. The model is composed of five
 228 compartments describing the five possible states of labeled carbon: dissolved (D), adsorbed (A),
 229 sequestered (S) state, or (following metabolism) carbon dioxide (CO₂) and living and dead biomass (X and
 230 X_{dead}). The model was implemented as a set of ordinary differential equations (ODEs) in MATLAB. The
 231 model was also successfully implemented in Microsoft Excel and produced equal results. The calculated
 232 sum of living and dead biomass was considered to be bioNER, and the sum of sequestered fraction and
 233 bioNER was compared to measured total NER. No kinetic data were available to separately simulate the
 234 formation of type II NER. Hence, any type II NER formed in the experiments were considered to be
 235 included in the sequestered compartment of the model. The calculated labeled carbon in the dissolved
 236 and adsorbed compartment was compared to the measured extractable labeled carbon.

237

238 *Calculation and fitting of input parameters*

239 Input parameters for the simulations were derived as follows. The initial amount of ¹³C was assumed to be
 240 distributed between the dissolved and solid phase according to the soil-water distribution coefficient (K_d).
 241 The sequestered fraction was assumed to be equal to the NER measured at $t = 0$ and corrected by the
 242 reported recovery. $NER(t = 0)$ was subtracted from the calculated ¹³C present in the solid phase to yield
 243 the adsorbed fraction.

244

245 We adjusted the input parameters step-wise, similar to Rein *et al.*³⁶:

246 Step i) Yield: Calculation with the MTB method.

247 Step ii) Death rate b : Towards the end of the experiments, the substrate is used up and the microbes
 248 decline. Then

$$249 \quad \ln \left(\frac{X(t)}{X(0)} \right)_{final\ phase} \approx -b t \quad (\text{eq. 18})$$

250 where X here is the measured concentration of microbial biomass; it is calculated from the measured
 251 PLFA times a factor of 20 (5% content of PLFA in native biomass).

252 Step iii) Growth: During the initial growth phase the microbial growth can be described as

$$253 \quad \ln \left(\frac{X(t)}{X(0)} \right)_{initial\ phase} = (\mu - b)t \quad (\text{eq. 19})$$

254 The resulting growth rate μ at given time t is used to estimate μ_{max} (SI 3.2).

255 Step iv) Half-saturation constant: For 2,4-D, a literature value for the ratio μ_{max}/K_M is given in Tuxen *et*
256 *al.*⁴⁴. For ibuprofen, K_M was fitted using the CO₂ development as criterion.

257 Step v) Initial degrader biomass: $X(0)$ was adjusted to fit the peak biomass concentration and the lag
258 phase. The sum of root mean square errors (RMSE) was used to describe the “goodness-of-fit” (SI S3.3).

259
260 During the model calibration against the 2,4-D data we found that the sequestration (slow adsorption,
261 leading to NER) of the labeled carbon of 2,4-D was better described by using the K_{OC} of 2,4-
262 dichlorophenol (2,4-DCP) instead of the K_{OC} of the parent compound 2,4-D. 2,4-DCP is the transformation
263 product of 2,4-D and has a K_{OC} much higher than 2,4-D. For the rapid adsorption (part of the extractable
264 ¹³C), the K_{OC} of 2,4-D was kept. It is well known that chlorinated phenols tend to form abiotic NER,¹² thus
265 the better fit of the 2,4-DCP K_{OC} may provide an indication for NER type II bonding via covalent bonds
266 triggered by oxidative coupling. The most appropriate way to accommodate this change would be the
267 inclusion of step-wise degradation (e.g., 2,4-D to 2,4-DCP to CO₂), but this increase in model complexity
268 would not be justified by the available data.

269

270 *Uncertainty analysis and parameter identification*

271 Aside pre-calculation, two optimization routines were used for calibration of v_{max} , K_{OC} , $X(0)$, Y , and K_M
272 (only for ibuprofen). The Pattern Search optimization function is an algorithm that finds local minima from
273 a mesh around the initial values and stops when the optimization function cannot be further minimized.⁴⁵

274 The Bayesian optimization method DiffeRential Evolution Adaptive Metropolis algorithm (DREAM_(ZS))^{46,47}
275 uses the Bayesian framework and also allows for the assessment of uncertainties related to the parameter
276 estimates and the model predictions. The Bayesian optimization was done with and without Y as a pre-
277 calculated parameter in order to assess the effect of its inclusion on the parameter estimation and on the
278 uncertainty of the model predictions. For further details on the parameter settings see the SI (S4).

279

280 **Results**

281 *Comparison of yield estimates*

282 Table 1 and Table 2 list the observed and estimated yields of substrates relevant to biotechnology (unit g
283 C g⁻¹ C) and of chemicals of environmental concern (unit g microbial biomass dw g⁻¹ substrate). The
284 biotechnological substrates are easily degradable compounds for which experimental yield data are
285 available.⁴⁸ Both the TEEM2²³ and the presented MTB yield estimation methods give relatively close
286 estimates with a mean absolute error (MAE) of less than 0.1 g C g⁻¹ C. Few experimental yield data are
287 available for chemicals of environmental concern. The estimates are less accurate, with the highest
288 deviation for linuron, which had a very low measured yield.⁴² Despite its simplicity, the MTB method overall
289 gave results with lower deviation compared to TEEM2 for the chemicals of environmental concern.

290

291 <Table 1>

292 <Table 2>

293

294 *Dynamic model simulations*

295 Figure 1 shows the experimental and the simulation results for ¹³CO₂, extractable ¹³C (dissolved and
296 adsorbed) and non-extractable ¹³C (which is the sum of ¹³C-label sequestered and in living or dead
297 biomass). For both compounds, the model with pre-calculated input data is able to reasonably describe
298 the fate of ¹³C in the different compartments. However, CO₂ and NER are predicted to increase at an
299 earlier time point than observed. For 2,4-D, this can be seen already at the first data points, whereas for
300 ibuprofen it is evident after 28 days. Based on the *Michaelis-Menten* equation, it was assumed that the
301 formation of CO₂ and new biomass occurs as soon as the labeled compound is transformed. In reality,
302 internal storage of metabolites and HCO₃⁻ delays the release of CO₂. This may be overcome by the
303 introduction of new parameters; only, this would considerably increase model complexity which was not
304 desired. In the beginning of the simulation, most NER are sequestered, but towards the end of the
305 simulation, the NER originate mainly from living and dead biomass.

306 The experimentally determined extractable ¹³C-ibuprofen was declining within four weeks (Fig. 1b). The
307 extractable ¹³C-label had initially similar values but remained relatively high throughout the experiment.
308 After 90 days, 13.4% of the ¹³C was detected in the solvent-extractable portion, but only 0.5% was ¹³C-

309 ibuprofen (Girardi *et al.*²⁹, Table S10). This indicates rapid formation of transformation products and
310 incomplete mineralization with only little 2-hydroxy-ibuprofen (Table S10).

311
312 Figure 2 depicts the simulated living biomass (X), dead biomass (X_{dead}) and the sum of both. This is
313 compared to the measured ^{13}C in PLFA multiplied with a factor 20 as a marker for living biomass (5%
314 PLFA content), and to the measured ^{13}C -label in tAA. Living biomass contains about 50% proteins (amino
315 acids), hence also tAA multiplied with a factor of two was plotted. It can be seen that PLFA/0.05 and
316 tAA/0.5 as well as the simulated sum of X and X_{dead} are close until day 4 (2,4-D, Fig. 2a) or day 14
317 (ibuprofen, Fig. 2b), as long as the living biomass predominates. Later PLFA declines, which indicates a
318 decline of living biomass X . The dotted line in Figure 2 is the decay half-time ($\ln 2 / b$) after maximum
319 measured PLFA. The line indicates where >50% of microbes have died. From this point, the simulated
320 sum of X and X_{dead} is much closer to tAA than to tAA/0.5. In decaying microbes, labile constituents like
321 sugars and fatty acids are turned over first and the more stable amino acids (tAA) in proteins persist (also
322 see SI S2.13).^{39,40} Thus, towards the end of the simulation, the sum of X and X_{dead} is dominated by
323 proteins and should be compared to tAA and not to tAA/0.5.

324

325 *Calculated bioNER versus measured tAA*

326 The measured $^{13}\text{CO}_2$ release in the 2,4-D experiment was 57.6% of the initially applied ^{13}C (SI Table S9),
327 and the calculated yield of 2,4-D was $0.28 \text{ g } ^{13}\text{C g}^{-1} ^{13}\text{C}$. Using these values in the equation for the ratio of
328 biomass growth to CO_2 production (eq. 4) we calculated that 22% of the applied ^{13}C -label was fixed in the
329 biomass. The measured tAA was at 23.3% (SI Table S9). For ibuprofen, with a measured $^{13}\text{CO}_2$ -release
330 of 45.2% (Table S10) and a calculated yield of $0.43 \text{ g } ^{13}\text{C g}^{-1} ^{13}\text{C}$, the calculated ^{13}C -label in biomass was
331 34% (measured tAA: 28.4%). In the case of these two experiments, the measured ^{13}C -label within amino
332 acids (tAA) was remarkably constant towards the end of the experimental period, and there was no need
333 to consider turnover of dead biomass. The calculated ^{13}C -label fixed in bioNER with eq. 5 was 9.4% (2,4-
334 D) and 12.4% (ibuprofen). Once the fraction of bioNER is known from Y and CO_2 , the potentially
335 remobilisable NER type I and type II can be quantified from the total NER. In the PBT/vPvB assessment of

336 chemicals, the bioNER fraction can be subtracted from the total NER and counted as degraded. In a
337 follow-up study, we used this method to estimate the bioNER for 40 chemicals of environmental concern.⁴⁹

338

339 <Figure 1>

340 <Figure 2>

341

342 Discussion

343 *Yield estimates*

344 We presented the new MTB approach for estimation of microbial biomass yields. Considering the
345 variability of the experimental data, this approach showed fairly similar deviations from experimental yield
346 data in comparison to the more advanced and widely applied TEEM2 approach²³, without the need for
347 specific information about the catabolic pathway, primary oxidation processes or N sources. For
348 environmentally relevant chemicals and pesticides the deviation of the experimental yields is even lower
349 than estimated with the MTB . The MTB approach can be applied for many tasks, e.g. yield assessment in
350 biological wastewater treatment or maximum transfer of labeled carbon into microbial biomass and
351 bioNER assessment, as shown with the simulations. Yield estimates can thus contribute to an improved
352 risk assessment of environmentally relevant chemicals. The method may be added as a module in
353 biodegradation databases (like EAWAG-BBD/PPS <http://eawag-bbd.ethz.ch/>, KEGG
354 <http://www.genome.jp/>) and QSAR approaches (like ChemProp www.ufz.de/ecochem/chemprop or EPI
355 suite <https://www.epa.gov/tsca-screening-tools/epi-suitetm-estimation-program-interface>). Combined with
356 the unified model for sorption and biodegradation (Kästner *et al.*¹², Rein *et al.*³⁶, and this study) the MTB
357 yield estimation method can be used for modeling the entire turnover process of a chemical in the
358 environment.

359

360 *Comparison to other findings*

361 Yield estimations are rarely applied to chemicals of environmental concern. One exception is the study of
362 Helbling *et al.*⁴² with linuron and carbofuran. The estimated theoretical yields for carbofuran are 0.51 g g⁻¹
363 (MTB method) or 0.59 g g⁻¹ (TEEM2 method) and 0.41 g C g⁻¹ C = 0.50 g g⁻¹ with the related adapted
364 TEEM1 method.⁴² The experimentally determined yield of carbofuran was 0.52 g g⁻¹ (0.42 g C g⁻¹ C). The
365 experimental yields obtained for linuron in the Helbling *et al.*⁴² study were very low (0.06 g C g⁻¹ C = 0.05 g
366 g⁻¹) despite a theoretical yield similar to carbofuran (0.40 g C g⁻¹ C = 0.33 g g⁻¹). Maximum growth rates
367 μ_{max} were determined to be 7.8 d⁻¹ (carbofuran) and 1.3 d⁻¹ (linuron), corresponding to a v_{max} = 15.1 and
368 26.4 g (g d)⁻¹, respectively.

369 Kinetic parameters and yields of polycyclic aromatic hydrocarbons PAH have been determined in several
370 studies. Wick *et al.*⁵⁰ grew *Mycobacterium* sp. LB501T on solid anthracene and obtained yields between
371 0.158 and 0.196 g g⁻¹ and v_{max} of 18.4 g (g d)⁻¹. Adam *et al.*⁵¹ found for the growth of three degrader
372 strains on phenanthrene the same yield of 0.21 g g⁻¹, with v_{max} from 12 to 18 g (g d)⁻¹. Rein *et al.*³⁶ tested
373 growth of *Mycobacterium* sp. on phenanthrene and pyrene and found yields from 0.20 to 0.32 g g⁻¹ and
374 v_{max} from 8 to 10 g (g d)⁻¹. Toräng *et al.*⁵² estimated a yield of about 0.3 g C g⁻¹ C for the degradation of U-
375 ring-labeled phenoxy-acetic acids (MCPD and 2,4-D) using the ¹⁴C-MPN (most probable number).⁵³

376 Most of the experimental yields (Table 2) are lower while v_{max} -values and growth rates are higher than
377 those obtained here (Table 3), and there could be several reasons for this: In these studies, known pure
378 degrader strains were tested under optimal nutrient conditions, which explains the faster growth and the
379 lower K_M -values compared to the studies simulated here, in which natural microbial communities were
380 used according to the OECD guidelines. Compound turnover and the related yields in experiments with
381 natural inoculum and multiple substrates may be lower than single-strain/single-substrate experiments due
382 to the enrichment of metabolites (incomplete mineralization) or to the use of multiple carbon sources
383 derived from dissolved organic carbon or SOM.^{31,42}

384

385 *Uncertainty analysis and parameter identification*

386 The pre-calculated model input parameters were compared to those fitted by the DREAM_(ZS) and the
387 Pattern Search algorithms (Table 3). For 2,4-D, the fitted yields are higher than the pre-calculated one.

388 Conversely, fitted yields for ibuprofen are substantially lower than the pre-calculated theoretical yield,
389 which may be again an indication of incomplete mineralization of ibuprofen. Both v_{\max} and K_M derived by
390 the DREAM_(ZS) algorithm are clearly higher than the pre-calculated values and those derived by the
391 Pattern Search algorithm, and this affects also the μ_{\max} -values. However, the ratio between v_{\max} and K_M ,
392 which is effectively determining metabolism (eq. 1), is for 2,4-D comparable amongst all four methods. For
393 ibuprofen this ratio is higher for the DREAM_(ZS) algorithm but within a factor of two of the values derived by
394 the other methods. The DREAM_(ZS) algorithm returned K_{OC} -values for the 2,4-D simulation that are very
395 close to the K_{OC} of 2,4-DCP. The value found with the Pattern Search algorithm is in between the K_{OC} -
396 values of 2,4-D and 2,4-DCP. A large disagreement between fitted and pre-calculated values is observed
397 for the K_{OC} of ibuprofen, where the pre-calculated value was obtained by a regression equation⁵⁴. Without
398 exception, the pre-calculated parameters are within the 95% credibility interval given by the DREAM_(ZS)
399 method. This gives additional confidence to the identified system kinetics.

400 A simultaneous fit of all parameters, as it is often done (for example, Brimo *et al.*⁵⁵), can produce a better
401 fit to experimental data. This was also the case in our simulations, where the RMSE of the simulated
402 results was lower when the input parameters were fitted (Table 3). Still, estimating the yield with an
403 independent method showed some advantages for the simulation. The parameter identifiability improved,
404 as can be seen from a decrease of the correlation between the fit parameters (Table S7). Using the
405 criteria of Frutiger *et al.*⁵⁶ ($r < 0.7$, $\sigma/\mu < 0.5$), all parameters were identifiable via model calibration to the
406 2,4-D data. For ibuprofen, only Y , K_{OC} , and K_M but not v_{\max} and $X(0)$ were identifiable (Table S8) (details in
407 SI section S4). The largest effect was seen on the uncertainty of the prediction: omitting Y from the fit
408 procedure greatly reduced the uncertainty in the model predictions, as shown by the width of the 95th-
409 percentile credibility interval (Figures S5-S8), in particular for NER and X . Importantly, as we showed in
410 this study, the knowledge of the yield gives insight into the degradation processes. It is now possible to
411 elucidate the nature of non-extractable residues by a combination of novel analytics, basic principles, and
412 dynamic simulation.

413

414 **<Table 3>**

415

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420

421 The dynamic degradation model with description is available in a public version at

422 <http://www.magicpah.org/links/> or <http://homepage.env.dtu.dk/stt/>.

423 Supporting Information available comprises: more detailed equations and parameter of the modeling
424 approach. This information is available free of charge via the Internet at <http://pubs.acs.org/>

425

426 The **MTB theoretical yield tool** is available as excel or Python code on request to the first author.

427

428

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573

574 **Tables**

575 **Table 1.** Comparison of yield estimates (g biomass carbon g⁻¹ substrate carbon, g C g⁻¹ C) using the
 576 TEEM2²³ and the MTB methods for biotechnological substrates. AE is absolute error and MAE is mean
 577 absolute error. Measured yields are taken from ref. 48.

| Biotech. substrates | Measured | TEEM2 | AE | MTB | AE |
|----------------------------|-----------------|--------------|---------------|------------|---------------|
| Acetate ⁻ | 0.41 | 0.40 | 0.01 | 0.47 | 0.05 |
| Citrate ³⁻ | 0.365 | 0.34 | 0.025 | 0.29 | 0.075 |
| Ethanol | 0.53 | 0.67 | 0.14 | 0.60 | 0.07 |
| Formaldehyde | 0.47 | 0.51 | 0.04 | 0.58 | 0.11 |
| Glucose | 0.61 | 0.48 | 0.13 | 0.61 | 0.0 |
| Glycerol | 0.67 | 0.55 | 0.12 | 0.62 | 0.05 |
| Glyoxylate | 0.22 | 0.27 | 0.05 | 0.27 | 0.05 |
| Methanol | 0.54 | 0.56 | 0.02 | 0.66 | 0.12 |
| Propionate ⁻ | 0.48 | 0.47 | 0.01 | 0.50 | 0.02 |
| Pyruvate ⁻ | 0.32 | 0.39 | 0.07 | 0.39 | 0.07 |
| MAE | | | 0.0615 | | 0.0615 |

578

579

580 **Table 2.** Comparison of yield estimates (g microbial biomass dw g⁻¹ substrate, g g⁻¹) using the TEEM2²³
 581 and the MTB methods for chemicals of environmental concern. AE is absolute error and MAE is mean
 582 absolute error.

| Environmental chemicals | Observed | TEEM2 | AE | MTB | AE | Reference for observed Y |
|---|------------|-------|--------------|------|--------------|---------------------------|
| 2,4-D (¹² C) | 0.25 | 0.39 | 0.14 | 0.23 | 0.02 | ⁵⁷ |
| 2,4-D (¹³ C ring-labeled) | 0.18; 0.25 | | | | | This study; ⁵² |
| 2,4-DCP | 0.30 | 0.41 | 0.11 | 0.21 | 0.09 | ⁵⁷ |
| Benzene | 0.71 | 0.84 | 0.13 | 0.65 | 0.06 | ⁵⁸ |
| Carbofuran | 0.52 | 0.59 | 0.07 | 0.51 | 0.01 | ⁴² |
| Carbon tetrachloride | | 0 | | 0 | | Persistent |
| DDT | | 0.42 | | 0.25 | | |
| Ibuprofen (¹² C) | 0.43 | 0.61 | 0.18 | 0.62 | 0.19 | This study |
| Ibuprofen (¹³ C ring-labeled) | 0.39 | | | | | ²⁸ |
| Linuron | 0.05 | 0.33 | 0.28 | 0.32 | 0.27 | ⁴² |
| Nitrilotriacetic acid | 0.23 | 0.23 | 0.00 | 0.27 | 0.04 | ⁴³ |
| Pentachlorophenol | | 0.19 | | 0 | | Persistent ¹⁶ |
| Phenanthrene | 0.21 | 0.82 | 0.61 | 0.53 | 0.32 | ⁵¹ |
| Pyrene | 0.32 | 0.54 | 0.22 | 0.44 | 0.12 | ³⁶ |
| Trichloroethene | | 0.16 | | 0.11 | | Persistent |
| Toluene | 0.71 | 0.86 | 0.15 | 0.69 | 0.02 | ⁵⁸ |
| MAE | | | 0.189 | | 0.114 | |

583
 584

585 **Table 3.** Input and fit parameters used for the simulation of degradation experiments of 2,4-D and
 586 ibuprofen described in Nowak *et al.*^{27,28} and Girardi *et al.*²⁹. Parameter values highlighted in bold were not
 587 fitted but pre-calculated.

| Parameter | unit | Manual w/ pre-estimated yield | Pattern Search w/o pre- estimated yield | DREAM w/ pre- estimated yield (95% credibility interval) | DREAM w/o pre-estimated yield (95% credibility interval) |
|-------------------------------|---|---|---|---|---|
| 2,4-D | | | | | |
| Y | $\text{g } ^{13}\text{C}_{\text{biomass}} (\text{g } ^{13}\text{C}_{\text{substrate}})^{-1}$ | 0.28^a | 0.36 | 0.28^a | 0.31 (0.21; 0.52) |
| b | d^{-1} | 0.05^b | 0.05^b | 0.05^b | 0.05^b |
| μ_{max} | $\text{g } ^{13}\text{C}_{\text{biomass}} (\text{g } ^{13}\text{C}_{\text{substrate}} \text{d})^{-1}$ | 1.1 ^c | 1.43 | 1.61 (0.38; 2.72) | 1.73 (0.40; 3.85) |
| $V_{\text{max}}/K_{\text{M}}$ | $\text{m}^{-3} (\text{g } ^{13}\text{C} \text{d})^{-1}$ | 2.72 ^d | 2.08 ^d | 2.72 ^d | 2.47 ^d (1.45; 3.61) |
| $X(0)$ | g m^{-3} | 0.172 | 0.28 | 0.87 (0.16; 1.34) | 0.88 (0.16; 1.4) |
| K_{OC} | L kg^{-1} | 2,4-D: 71.4^e ; 2,4-DCP: 689^f | 300 | 655 (218; 977) | 668 (506; 836) |
| sum | $\text{g } ^{13}\text{C m}^{-3}$ | 5.57 | 1.56 | 2.17 | 2.13 |
| RMSE ^g | | | | | |
| Ibuprofen | | | | | |
| Y | $\text{g } ^{13}\text{C}_{\text{biomass}} (\text{g } ^{13}\text{C}_{\text{substrate}})^{-1}$ | 0.43^a | 0.28 | 0.43^a | 0.32 (0.11; 0.53) |
| b | d^{-1} | 0.03^b | 0.03^b | 0.03^b | 0.03^b |
| μ_{max} | $\text{g } ^{13}\text{C}_{\text{biomass}} (\text{g } ^{13}\text{C}_{\text{substrate}} \text{d})^{-1}$ | 0.50 ^c | 0.28 | 1.41 (0.18; 4.1) | 1.34 (0.12; 3.85) |

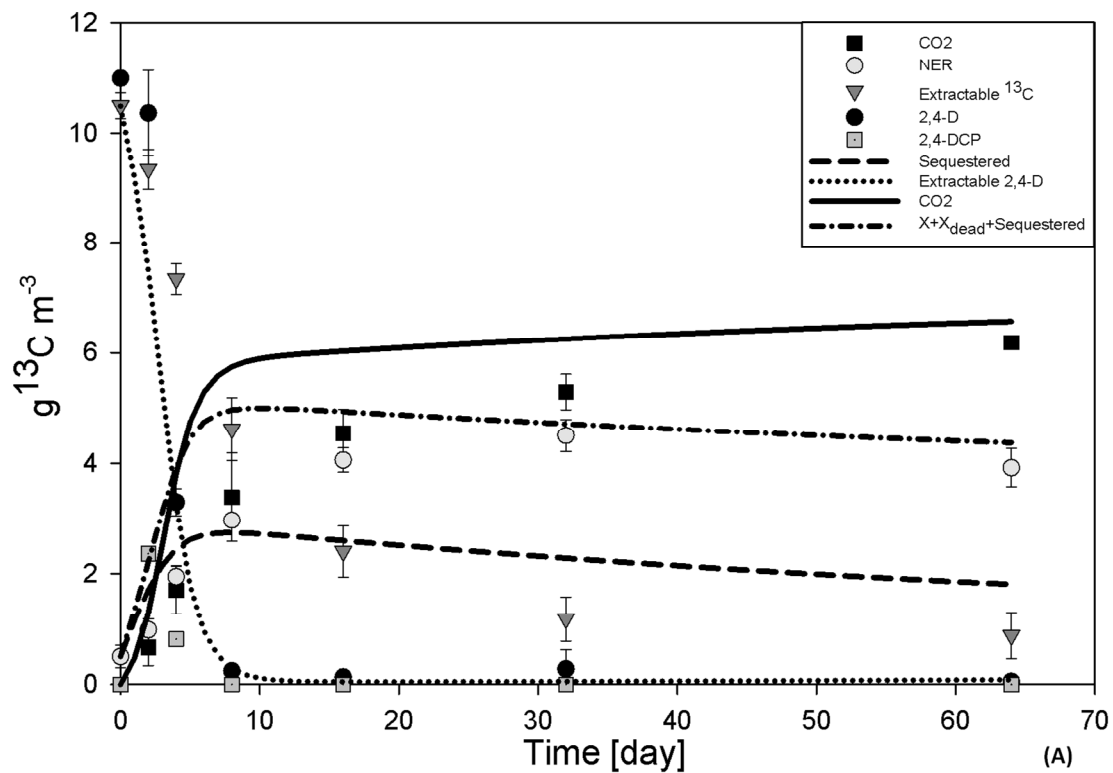
| | | | | | |
|-------------------------------|---|------------------|------|------------------|------------------|
| | $^{13}\text{C}_{\text{substrate}} \text{ d}^{-1}$ | | | | 3.7) |
| $X(0)$ | g m^{-3} | 0.069 | 1.2 | 0.69 (0.05; 1.3) | 0.62 (0.05, 1.3) |
| $v_{\text{max}}/K_{\text{M}}$ | $\text{m}^{-3} (\text{g}^{13}\text{C d})^{-1}$ | 0.39 | 0.35 | 0.51 (0.10; 2.4) | 0.67 (0.12; 2.7) |
| K_{OC} | L kg^{-1} | 108^{h} | 552 | 558 (83.8; 788) | 546 (99.3; 789) |
| sum | $\text{g}^{13}\text{C m}^{-3}$ | 4.62 | 1.89 | 2.86 | 2.97 |
| RMSE ^g | | | | | |

588 a: estimated with MTB method; conversion factor 2,4-D = 0.822 and ibuprofen = 1.41 for conversion to g
589 microbial biomass dw per g substrate; b: from slope of $\ln X$ at the end of the experiment; c: from slope of
590 $\ln X$ in the initial growth phase; d: K_{M} of 2,4-D was calculated⁵²; e: K_{OC} of 2,4-D (estimated⁵⁴) was used for
591 rapid adsorption; f: K_{OC} of 2,4-DCP (estimated⁵⁴) was used for slow adsorption (sequestration); g
592 description of sum RMSE see SI S3.3; h: estimated⁵⁴
593

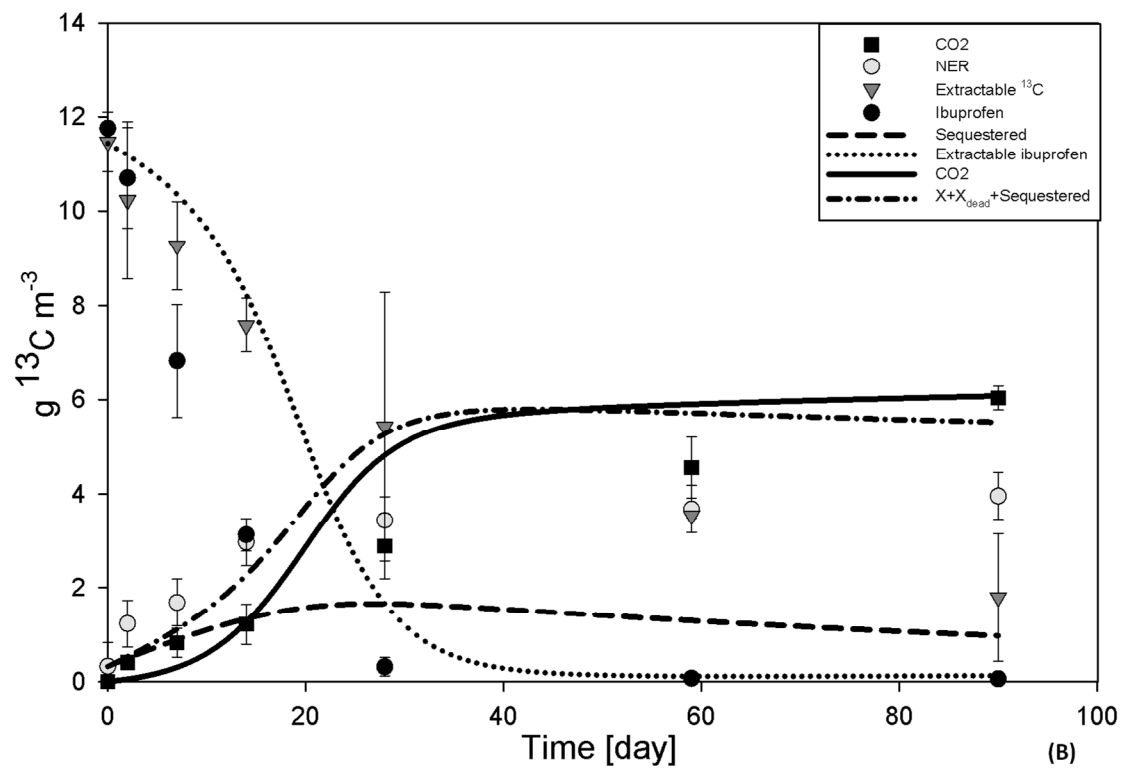
594 **Figure legends:**

595 **Figure 1.** Measured and simulated ^{13}C -label distribution. A) Top: $^{13}\text{C}_6$ -2,4-D, and B) bottom: $^{13}\text{C}_6$ -
596 ibuprofen. Symbols show measured data, curves show the simulated turnover. Symbols: CO_2 (black
597 square), NER (grey circle), extractable ^{13}C -label (dark grey triangle), and added compound, i.e., ibuprofen
598 or 2,4-D (black circle). Curves: Sequestered (black dashed (- -)), extractable compound (dark grey (-)),
599 CO_2 (black (-)), and living biomass + dead biomass + sequestered compound (=NER) (light grey). Error
600 bars show the standard deviation of the measurements as reported by Girardi *et al.*²⁹ and Nowak *et*
601 *al.*^{27,28}.

602
603
604 **Figure 2.** Simulations results for the growth of biomass. A) Top: $^{13}\text{C}_6$ -2,4-D and B) bottom: $^{13}\text{C}_6$ -ibuprofen.
605 Symbols show measured data, curves show simulation of the formation of living and dead biomass.
606 Symbols: Phospholipid fatty acids (PLFA; black circles), total amino acids (tAA; dark grey squares), and
607 total amino acids multiplied with a factor of two to yield total dead and alive biomass (empty squares).
608 Curves: Concentration of living biomass X (black), concentration of dead biomass X_{dead} (light grey), and
609 concentration of living and dead biomass $X + X_{\text{dead}}$ (dark grey). Error bars show the standard deviation of
610 the measurements as reported by Nowak *et al.*^{27,28}. The dotted vertical line is the halftime of decay ($\ln 2 /$
611 b) after maximum measured PLFA and indicates were $> 50\%$ of tAA is necromass.



612

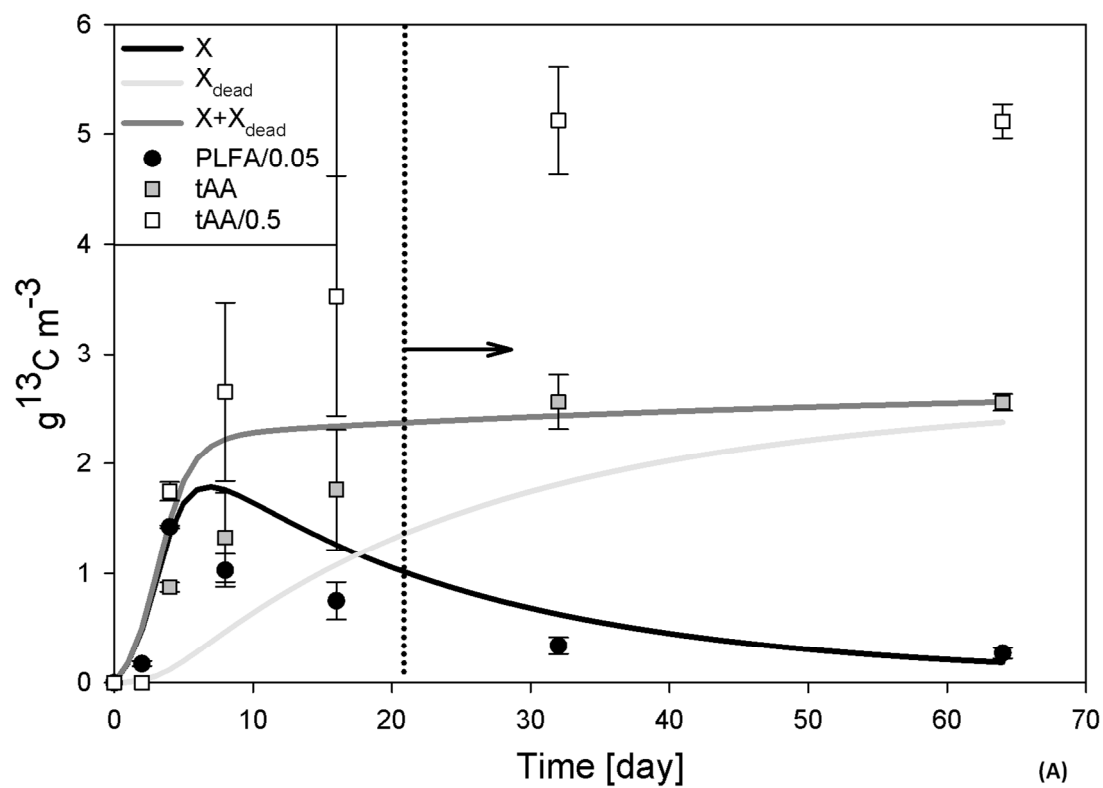


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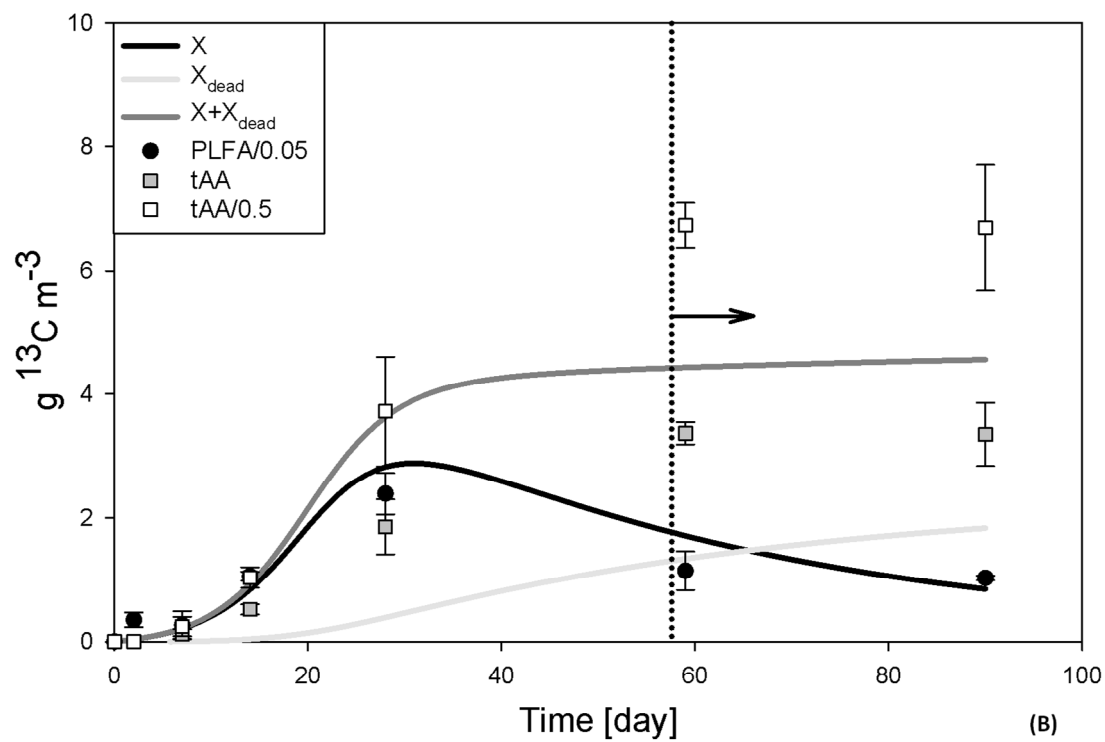
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615

Figure 1



616



617

618

Figure 2