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1	Unbound fractions of PFAS in Human and Rodent Tissues: Rat liver a
2	suitable proxy for evaluating emerging PFAS?
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18 Abstract

19 Adverse health effects associated with exposures to per- and polyfluoroalkyl substances (PFAS) are a concern for public health and are driven by their elimination half-lives and 20 21 accumulation in specific tissues. However, data on PFAS binding in human tissues are limited. 22 Accumulation of PFAS in human tissues has been linked to interactions with specific proteins and 23 lipids in target organs. Additional data on PFAS binding and unbound fractions ($f_{unbound}$) in whole 24 human tissues are urgently needed. Here we address this gap by using rapid equilibrium dialysis to measure the binding and $f_{unbound}$ of 16 PFAS with 3 to 13 perfluorinated carbon atoms (η_{pfc} = 3-25 13) and several functional head-groups in human liver, lung, kidney, heart, and brain tissue. We 26 27 compare results to mouse (C57BL/6 and CD-1) and rat tissues. Results show that funbound 28 decreases with increasing fluorinated carbon chain length and hydrophobicity. Among human tissues, PFAS binding was generally greatest in brain>liver≈kidneys≈heart>lungs. A correlation 29 analysis among human and rodent tissues identified rat liver as suitable surrogate for predicting 30 f_{unbound} for PFAS in human tissues (R² ≥0.98). The f_{unbound} data resulting from this work and the rat 31 32 liver prediction method offer input parameters and tools for toxicokinetic models for legacy and 33 emerging PFAS.

34 <u>Synopsis:</u> Similar PFAS binding in human and rodent tissues suggests rat liver is a useful model
 35 tissue for predicting PFAS accumulation and tissue distribution in humans.

36 **Keywords:** PFAS, unbound fractions, tissue binding, species correlation, equilibrium dialysis

37 TOC Art



39 1. Introduction

Per- and polyfluorinated alkyl substances (PFAS) are a diverse class of anthropogenic 40 chemicals with a wide range of commercial and industrial applications.¹ Their extensive use has 41 42 resulted in detectable concentrations in drinking water, ambient air, animals, and in virtually all humans.¹⁻¹³ PFAS exposures are associated with adverse health outcomes, such as thyroid 43 hormone disruption, decreased response to vaccination, ¹⁴ dyslipidemia, ¹⁵ and hepatotoxicity. ¹⁶ 44 Slow elimination and strong tissue accumulation of PFAS are typically associated with increased 45 toxicity.¹⁷ While hundreds of PFAS have been detected in the environment, tissue-specific 46 accumulation has been tested only for a handful of PFAS.¹⁸ Accumulation and slow elimination 47 of PFAS in humans have been associated with binding to isolated biomolecules; ¹⁹⁻²¹ however, 48 they have not yet been linked to PFAS binding to whole tissues. Data and predictive tools for 49 50 characterizing PFAS binding in whole human tissues are needed and are the focus of this work.

51 PFAS with long fluorinated carbon chains ($\eta_{pfc} \ge 7$), such as perfluorooctanoic acid (PFOA, η_{pfc} = 7) and perfluorooctane sulfonic acid (PFOS, η_{pfc} = 8), show strong accumulation in humans 52 and have elimination half-lives that range between 3-6 years.²²⁻²⁴ In contrast, short-chain PFAS 53 like perfluorobutanoic acid (PFBS, η_{pfc} = 3) are eliminated by humans in weeks.²⁵ After absorption 54 in the gastrointestinal tract, PFAS pass through liver and enter the circulatory system. ^{21,26-28} For 55 56 PFAS with $\eta_{pfc} \ge 7$, binding to proteins and phospholipids are thought to determine elimination halflives.²⁹⁻³¹ For PFAS with η_{pfc} <6, membrane transporters, enterohepatic recirculation, and renal 57 tubular reabsorption may play a more important role in tissue distribution.^{31,32} Binding of PFAS to 58 59 tissues is driven by their physicochemical properties (e.g., η_{pfc} , hydrophobicity, pKa), the 60 macromolecular composition of the tissue (e.g., structural proteins, phospholipids), and abundantly expressed functional proteins (e.g., fatty acid binding proteins and albumin). ^{33,34} An 61 increasing number of structurally diverse PFAS means there is a need to systematically 62 63 characterize PFAS binding to human tissues and identify the physicochemical properties that drive tissue accumulation. However, PFAS binding to human tissues beyond plasma have not yetbeen measured.

The unbound fraction of chemicals ($f_{unbound}$) in tissues is typically the amount that is 66 available for inter-tissue transport and interacts with molecular targets that elicit toxic effects.³⁵ 67 68 $f_{unbound}$ is inversely correlated to the tissue binding affinity, which can be quantified by the tissuewater partition coefficient ($K_{tissue/w}$).³¹ PFAS that are bound to tissue components (e.g., proteins, 69 70 phospholipids) are considered unavailable for renal and hepatic clearance until being desorbed as unbound PFAS are eliminated, ³⁶⁻³⁸ thereby prolonging elimination half-lives. At steady-state, 71 unbound concentrations ($C_{unbound}$) are expected to be similar among tissues, assuming negligible 72 influence of membrane transporter-mediated flux.³⁹ $f_{unbound}$ and $K_{tissue/w}$ can be used to predict 73 tissue-specific accumulation and elimination half-lives. Comprehensive studies on experimental 74 f_{unbound} and $K_{\text{tissue/w}}$ in whole human tissues are urgently needed to assess PFAS accumulation 75 and tissue distribution in humans. 76

77 Previous studies measured PFAS binding to isolated macromolecules, such as proteins (bovine serum albumin [BSA] 40,41, human serum albumin [HSA] 42, fatty-acid binding protein 78 [FABP]⁴³) and phospholipids.⁴⁴ Partition coefficients derived from such studies have been used 79 to parameterize physiologically-based toxicokinetic (PBTK) models and simulate $f_{unbound}$ and the 80 tissue distribution of PFAS.^{19,20,29,31} Machine-learning and molecular dynamics simulations are *in* 81 silico substitutes to experimental derivation of $f_{unbound}$ and tissue binding.⁴⁵ While these 82 simulations provide mechanistic insights into the interaction of PFAS with single biomolecules, ³⁸ 83 84 they have rarely been evaluated against experimental funbound in whole tissues. Rodent tissues have been used to generate data on $f_{unbound}$ for diverse pharmaceuticals that were useful for 85 predicting $f_{unbound}$ in human tissues.⁴⁶⁻⁴⁹ Extending tissue partitioning measurements for PFAS in 86 87 rodents to humans might similarly be useful.

The main objective of this work was to evaluate the suitability of $f_{unbound}$ and $K_{tissue/w}$ measured in rodent model tissues for predicting PFAS binding in human tissues. Furthermore,

90 this is the first study to report unbound fractions in various human tissues. This study leverages a well-established framework for pharmaceuticals and a high-throughput equilibrium dialysis 91 method⁵⁰⁻⁵² to measure $f_{unbound}$ and $K_{tissue/w}$ of 16 PFAS with η_{pfc} 3–13 and different functional head 92 93 groups (carboxylic acids, sulfonic acids, and sulfonamides) in brain, heart, kidney, liver, and lung tissues of humans and rodents (mouse, rat). These tissues were selected because they have 94 been shown to accumulate PFAS in humans.^{19,21} Inter-species and -tissue correlation analyses 95 were used to evaluate how well various rodent tissues predicted $f_{unbound}$ in human tissues. 96 Regression modeling was applied to assess the relationship between physicochemical properties 97 of PFAS and their funbound in various tissues. Using our data, we discuss potential mechanisms 98 99 driving PFAS toxicokinetics in humans and the suitability of rodent surrogate tissues as a fast and 100 straightforward tool for predicting human PFAS tissue distribution.

101 **2. Materials and Methods**

102 **2.1 Chemicals and Reagents.**

The study PFAS included 11 perfluoroalkyl carboxylates, 3 perfluoroalkyl sulfonates, 1 103 104 perfluoroalkyl sulfonamide, and 1 fluorotelomer sulfonate. Physicochemical properties of PFAS ranged from η_{pfc} from 3 to 13, molecular weight (M_w) from 214.04 to 714.11 g/mol, pKa from -0.21 105 to 6.25.⁵³ and octanol-water distribution coefficients (log*D*, pH: 7.0) from 2.23 to 10.11 (Tables 106 107 S1 and S2). Standards for perfluorobutanoic acid (PFBA), perfluoropentanoic acid (PFPA), perfluorohexanoic acid (PFHxA), perfluoroheptanoic acid (PFHpA), perfluorooctanoic acid 108 109 (PFOA), perfluorononanoic acid (PFNA), perfluorodecanoic acid (PFDA), perfluoroundecanoic acid (PFUDA), perfluorododecanoic acid (PFDoDA), perfluorotridecanoic acid (PFTriDA), 110 111 (PFTeDA), perfluorohexane perfluorotetradecanoic acid sulfonic acid (PFHxS), 112 perfluorooctanesulfonamide (PFOSA), and 6:2 fluorotelomer sulfonate (6:2 FtS) were purchased 113 from AccuStandard. Standards for perfluorobutane sulfonic acid (PFBS) and perfluorooctane sulfonic acid (PFOS) were purchased from Sigma-Aldrich. Brain, heart, kidney, liver, and lung 114 115 from commonly used rat and mouse strains (Wistar Han rat, CD-1 and C57BL/6J mouse) were 116 obtained from BioIVT and the Pfizer Drug and Safety Research & Development (DSRD) tissue 117 bank and pooled by sex. All mouse and rat work was performed according to Pfizer Global Research policies and procedures. Human heart, kidney, liver, and lung were purchased from 118 BioIVT, and human brain was obtained from the NIH NeuroBioBank⁵⁴ as a pool of 2 male and 2 119 120 female donors. All human tissues were processed following University of Rhode Island and Pfizer 121 Institutional Review Board approved protocols. The equilibrium dialysis device (HTD96) and 122 cellulose membranes with a molecular weight cutoff of 12-14 kDa were obtained from HTDialysis, 123 LLC. All other reagents were purchased from Sigma-Aldrich unless otherwise specified.

124 **2.2 Tissue homogenization and preparation**

Tissues were processed according to previously described methods.⁴⁷ Briefly, frozen 125 tissues were completely thawed in a 37°C water bath. Small incisions were made to the tissues 126 127 to minimize blood contamination. Thawed tissues were soaked in Dulbecco's phosphate-buffered 128 saline (DPBS, pH: 7.4, ThermoFisher) for five minutes. The pH in the suspension was measured for all experiments. Tissue samples were subsequently dried with paper towel, weighed, and 129 130 diluted to either 1:5 or 1:10 (brain) using DPBS. Tissues were homogenized using a TH tissue homogenizer (Omni). In a second step, the tissues were homogenized further using a glass 131 dounce homogenizer (Wheaton). 132

133 2.3 Equilibrium Dialysis

The equilibrium dialysis methods used an HTD 96 micro equilibrium device for binding 134 studies that was similar to previously reported techniques. ⁵⁰⁻⁵² Briefly, DBPS buffer solution 135 containing PFAS was prepared. The receiver concentration was ~five-fold the estimated $f_{unbound}$ 136 137 concentration based on preliminary data using the same test setup or was assumed to be 0.001 138 (unitless) if no prior data were available. Porous membranes of the dialysis apparatus were pre-139 saturated twice with 250 µL of PFAS-spiked DBPS buffer for one hour and overnight (~18 h), respectively. Stock solutions of single PFAS in dimethylsulfoxide (DMSO) were prepared and 140 added to tissue homogenates yielding final concentrations of 5 µM (1% DMSO). This 141 142 concentration was chosen to optimize the LC-MS/MS signal while maintaining a PFAS-to-tissue homogenate ratio that avoids saturation effects observed for proteins like serum albumin. ^{37,55} 150 143 144 µL of PFAS-free DBPS were added to the receiver and 150 µL of PFAS-spiked tissue 145 homogenates were added to the donor side. Replicates (n=4) of PFAS: tissue combinations were 146 tested. The HTD was sealed with a gas-permeable membrane (Sigma) and shaken at 200 rpm at 147 5% CO₂/air, 75% relative humidity, and 37 °C for 18 h. For protein and tissue precipitation, 15 μL of buffer and 45 µL of tissue homogenate were transferred into a 96-deep well plate. To ensure 148

there were no interferences between wells during the LC-MS/MS run, an additional 45 μ L of blank matrix was added to the buffer side and 15 μ L of blank buffer to the matrix side in the sample plate to create matrix-matched samples. The samples were quenched with 200 μ L of cold acetonitrile containing tolbutamide (30 nM) and carbamazepine (12 nM) that were used as internal standards to monitor the performance of the LC-MS/MS. The plates were sealed, vortexed for 1 min, and centrifuged at 3000 rpm for 5 min. Supernatant was transferred to a new 96-well, evaporated under nitrogen gas, and analyzed by LC-MS/MS. *f*_{unbound} were calculated as

156 Diluted
$$f_{\text{unbound,d}} = \frac{\text{Receiver Area Ratio}}{\text{Donor Area Ratio}}$$
 (1)

157 Undiluted
$$f_{\text{unbound}} = \frac{1/\text{DF}}{\left(\left(\left(\frac{1}{f_{\text{unbound,d}}}\right) - 1\right) + \frac{1}{DF}\right)}$$
 (2)

where DF is the dilution factor and $f_{unbound,d}$ is the diluted $f_{unbound}$. All calculations are based on area ratios (analyte/internal standard) within the linear range of calibration curves. Stability and recovery of PFAS concentrations were tested in parallel for all experiments and were above quality control thresholds (>70% recovery) in all experiments for all test compounds. Tissues that were not spiked with PFAS were used as controls and showed no or minimal background levels of the target compounds that did not infer with the data analysis.

164 **2.4 LC-MS/MS Quantification**

For analysis, samples were reconstituted in HPLC grade water/acetonitrile 50/50 (v/v) and vortexed for 1 min. LC-MS/MS analyses were performed on a SCIEX Triple Quad 5500 mass spectrometer (SCIEX) equipped with Turbo Ion Spray interface. The full analytical method is described in Section S2 and Table S3 of the Supporting Information (SI).

169 2.5 Scalar Determination and Statistical Data Analysis

170 Multiple analyses were conducted to evaluate the relationships between $f_{unbound}$ across 171 tissues and species following previous studies.^{47,48} The variability in quadruplicate $f_{unbound}$ were

172 log normally distributed, justifying the use of a log transformed funbound for statistical analysis. 173 Geometric mean funbound values for each PFAS were analyzed across tissues (brain, heart, kidney, liver, lung) and species (human, rat, mouse). To determine if rodent tissues can predict PFAS 174 175 binding in human tissues, we tested if existing scaling factors that were derived for pharmaceuticals⁴⁷ can be used for PFAS. Scaling factors are drug-independent numerical 176 coefficients that account for species-species differences in tissue composition and binding affinity. 177 178 enabling the extrapolation of drug distribution data from animal models to humans. New scalars for brain tissue $f_{unbound}$ from rat liver for this PFAS set by minimizing sum of squared errors (SSE) 179 180 between observed and predicted values as

181 SSE =
$$\sum (\ln(\text{Predicted}) - \ln(\text{Observed}))^2$$
 (3)

Scatterplots on the log scale were used to visualize the strength of relationships between rat liver 182 183 $f_{unbound}$ and other species/tissue combinations. Mean absolute fold differences (MAFD) and Spearman's rank correlations were used to quantify the agreement and were calculated as the 184 185 exponentiated mean of absolute log funbound differences, equivalent to mean absolute error on the log scale. Plots and summary statistics were produced with and without scalars. To evaluate the 186 equivalence of geometric mean $f_{unbound}$ values across species and tissues, we employed the two 187 one-sided tests (TOST) for average bioequivalence.^{48,56,57} The TOST procedure tested if the true 188 189 geometric mean funbound differences between species/tissue pairs were within a 2-fold difference, 190 resulting in 19 total statistical comparisons. The p-values and confidence intervals were adjusted using the Bonferroni adjustment to maintain a family-wise error rate of 0.05 and corresponding 191 confidence intervals.⁵⁸ This test was performed before and after the application of scalars. The 192 193 TOST procedure was also employed to assess equivalence for similar tissues between species. As all pairwise comparisons of species were assessed, the less conservative *l*-correction⁵⁹ was 194 used to adjust the intervals and maintain a family-wise error rate of 0.05. The data were analyzed 195 using R^{60} and the plots were created via the ggplot2 package.⁶¹ 196

197 2.6 Partial Least Squares Regression (PLSR)

Relationships between PFAS physicochemical properties and funbound in tissues were 198 199 investigated using a Partial Least Square Regression (PLSR) multivariate approach. We used funbound in the liver, brain, heart, kidney, and lung tissue to develop regression models. PLSR is a 200 201 multivariate approach that allows many responses (i.e., funbound in various tissues) to be incorporated in one regression model. PLSR performs a descriptor dimension reduction 202 203 procedure and constructs a set of latent components that maximize the total descriptors variance in the dataset to generate the model with the best predictive power for the target responses.⁶² 204 205 Using the PLSR approach, we developed four models based on the experimental $f_{unbound}$ results obtained from the four organisms: C57, human, mouse, and rat. A total of 20 physicochemical 206 207 properties were calculated from Molecular Operation Environment (MOE) software version 208 2018.01 (Chemical Computing Group ULC) and Dragon software version 6.0 (Talete srl). These property descriptors were combined to train and validate the PLSR models for the 16 study PFAS 209 210 using the standard Leave One Out Cross Validation (LOOCV). In the LOOCV procedure, a PFAS 211 was considered as a test set for model evaluation and the remaining 15 PFAS were used as 212 training set for developing the model. This process was repeated 16 times, such that each PFAS 213 was used as a test set once. Parameters for building the model (i.e., number of components) 214 were tuned using a grid search in the cross-validation process. The optimal number of 215 components was determined to be four for the model developed for humans, and three for the 216 models developed for C57, mouse and rat.

217 3.0 Results & Discussion

Unbound fractions of PFAS in human and rodent tissues. $f_{unbound}$ in human, rat, and mouse tissues generally decreased with molecular weight (M_w) and octanol-water distribution coefficient (log*D*) of the tested PFAS (Figs. 1, S1, S2). $f_{unbound}$ of the tested PFAS varied by six orders of magnitude (6.8 x 10⁻⁶ to 0.3), with six of the 16 PFAS being >99.999% bound. The SI contains all data on $f_{unbound}$ and coefficients of variation (Tables S4-S7). Higher molecular weight (M_w) PFAS with longer perfluorinated carbon chains and higher log*D* have high tissue binding affinity, resulting in lower $f_{unbound}$ than the shorter-chain, low M_w PFAS with low log*D*.



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Figure 1. A. *f*_{unbound} in human ("H") and rat ("R") liver, brain, heart, kidney, and lung tissue plotted against the molecular weight (g/mol) of the 16 study PFAS. B. Tissue/water partition coefficients

of the 16 study PFAS in human liver, brain, heart, kidney, and lung tissue.

229 We observed a strong relationship between M_w and $f_{unbound}$. For example, PFTeDA (M_w : 714) exhibited a low $f_{unbound}$ in human liver tissue (6.8 x 10⁻⁶), in contrast to PFBA (M_w: 214), which 230 showed six orders of magnitude higher $f_{unbound}$ of 0.3. Both PFAS are carboxylic acids, but the 231 232 increase in fluorinated carbon chain length and lipophilicity resulted in a 45,000x stronger binding 233 of PFTeDA (η_{pfc} = 14; log*D*: 10.11) than PFBA (η_{pfc} = 3; log*D*: 2.2). A similar trend for binding to 234 serum albumin and globulins was recently reported, where carboxylates showed increased binding as M_w increased up until a plateau at a $Mw \approx 500-600$ g/mol.²⁹ Interestingly, this plateau 235 was not observed in this study (Fig. 1). Past work suggested phospholipid concentrations drive 236 accumulation of $\eta_{ofc} \ge 10$ PFAS in the brains of marine mammals.⁶³ Accordingly, binding of PFAS 237 to isolated phospholipids increased linearly up until PFDA, 30,64 indicating $f_{unbound}$ of long-chain 238 239 PFAS to be driven by phospholipid binding more than by protein binding.

Fig. 1B shows the $f_{unbound}$ data converted into $K_{tissue/w}$ values to facilitate comparisons to human tissues. $K_{tissue/w}$ values were highest in brain tissues for 14 of the 16 PFAS. This potentially reflects the high protein and membrane lipid content of human brain tissue, ⁶⁵ which are known to strongly bind long-chain PFAS. ^{30,31} Comparable binding among human liver, kidney, and heart tissues was observed, with small variations among the PFAS tested. Binding in lung tissue was relatively low, especially for $\eta_{pfc} = 10-13$ carboxylic acids, which may reflect its low phospholipid content. ⁶⁵

Strong correlation between PFAS tissue binding in humans and rodents. Pairwise comparisons for all tissues and species demonstrated a strong association between logtransformed $f_{unbound}$ of PFAS between all tissues and species combinations (Figs. 2, S3, R² ≥0.97). These results suggest humans and rodents exhibit similar PFAS tissue binding for the brain, heart, lung, liver, and kidney. Spearman Rank Correlations among all the comparisons ranged from 0.96 to 1 (Table S8). These data demonstrate approximate inter-matrix agreement between $f_{unbound}$ of PFAS across the different species and tissue types. A Two One-Sided Tests (TOST) procedure for multiple comparisons was conducted within tissue pairwise species comparisons $f_{unbound}$ values for all tissues (Fig. S4). A 2-fold difference threshold analysis was employed to determine equivalence. ^{48,50} Of note, the highest concordance in log-transformed $f_{unbound}$ between human and rodent species was found for human tissues and rat liver (Fig. S5, R² ≥0.98), suggesting the utility of rat liver as a model tissue for estimating the binding of PFAS in human tissues.



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Figure 2. Pairwise comparison of human tissue *f*_{unbound} versus rat tissue *f*_{unbound} for brain, heart, kidney, liver, and lung. Red dashed lines indicate 1:1 correlations.

Rat liver as a surrogate for predicting tissue binding of PFAS in humans. Previous work on pharmaceuticals has used rat liver $f_{unbound}$ as a surrogate for estimating $f_{unbound}$ in human heart, liver, lung, kidney, and brain tissues.⁴⁷ In this work, $f_{unbound}$ was not found to depend on 266 species and application of scaling factors for any differences allowed robust predictions of human tissue binding. ⁶⁶ In this study, we adapted previously established scalars⁴⁷ to predict tissue 267 binding of PFAS in humans using rat liver as model tissue. Comparisons of rat liver funbound against 268 269 the funbound for multiple tissues and species, with and without scalars, are shown in Figs 3a and 3b 270 with 2-and 10-fold thresholds. Most comparison matrix pairs had a pre-scalar mean absolute fold difference (MAFD) below 2.0 except for human and rat lung showing an MAFD of 2.04 and 2.21 271 272 (Table S9), respectively. The MAFD after applying scalars against rat liver $f_{unbound}$ ranged from 1.19 to 1.57, indicating 2-fold equivalence across all data sets. The Spearman rank correlation of 273 rat liver against all combinations showed a high correlation with a range between 0.965 and 0.997 274 (Table S9). 275





Bonferroni adjusted TOST analysis across all species and tissues against rat liver $f_{unbound}$ values before and after applying scaling factors improved concordance between species (Fig. 3). 280 New scaling factors for brain tissue were estimated based on our PFAS data set (Table 1). The 281 agreement between rodent and human $f_{unbound}$, when applying scalars previously derived for pharmaceuticals, ⁴⁷ validates their use for extrapolating PFAS tissue binding across species. 282 283 Using the scalars, all tissue types were statistically 2-fold equivalent to rat liver, except brain (Fig. 3). Rat brain was concordant with rat liver for pharmaceuticals, ⁴⁷ this was not the case for our 284 PFAS data set. The higher binding for brain in this study may be due to the higher phospholipid 285 content of brain compared to other tissues.⁶⁷ For example, PFOSA (log*D*: 5.5) had a MAFD of 286 1.02 when comparing liver against brain for all species, whereas PFOS (logD: 6.8) had a MAFD 287 of 2.06. Although brain binding for all species was not within the 2-fold equivalency threshold, 288 data were within 3-fold. Overall, our study suggests equivalence of funbound values between all 289 groups against rat liver when scaling factors⁴⁷ are applied, suggesting rat liver can be used as a 290 291 surrogate for predicting PFAS tissue binding in future studies.





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Figure 3. Bonferroni adjusted Two One-Sided Tests (TOST) procedure analysis tissue $f_{unbound}$ equivalence test against rat liver $f_{unbound}$ with unscaled (A) and scaled (B) values.

Physicochemical properties drive tissue binding of PFAS. Results of this study show

296 a clear relationship between PFAS molecular weight and $f_{unbound}$, as well as log D and $f_{unbound}$ (Figs. 297 1, S1). As a next step, we applied Partial Least Squares Regression modeling to investigate the relationship between 20 physicochemical properties of PFAS and their funbound in tissues of the 298 299 four species/strains (human, rat, CD-1 mouse, and C57BL/6 mouse). The results demonstrated 300 strong associations between physicochemical properties and $f_{unbound}$ (Table S10). The correlation coefficients (R²) of leave one out cross validation are in the range from 0.95 to 0.99. By conducting 301 302 a descriptor analysis on the resulting models, we assessed the relative importance of individual physicochemical properties of PFAS to their overall tissue binding, based on their mean 303 304 coefficients in determining log-transformed $f_{unbound}$ (Table S11). Molecular weight and pKa were found to have the highest contributions to predict funbound among all 20 physicochemical descriptors 305 306 in resulted models (Table S11). All three lipophilicity descriptors, especially logD, contributed 307 positively to $f_{unbound}$ predictions (mean coefficients > 0). These results agree with previous studies 308 that demonstrated significant correlations between molecular weight, hydrophobicity, fluorinated 309 carbon chain length and the binding strength of PFAS to isolated tissue macromolecules like albumin^{29,38,41,42}, structural proteins, ³¹ and phospholipids. ^{30,44} 310

311 Significance of tissue binding to PFAS toxicokinetics. The generally strong and M_w-312 dependent funbound of PFAS in tissues suggest that tissue binding is essential for understanding 313 the distribution and the accumulation within tissues. Results showed binding of PFAS was 314 comparable between liver, kidney, and heart tissues. This suggests that specific binding proteins 315 may be less critical for tissue accumulation than previously postulated. For example, the liver is 316 enriched in liver-type fatty acid binding proteins (L-FABP), yet tissue binding to liver was comparable to other tissues (≤2-fold difference for 14/16 PFAS). This indicates that L-FABP may 317 be less important for tissue retention than previously suggested.⁶⁸ Data from this study suggest 318 that macromolecules (e.g., structural proteins, phospholipids) that have comparable contents 319 320 among tissues are the primary drivers of relatively small differences in $f_{unbound}$.

Comparability to animal studies and human cohorts. The long half-lives of long-chain 321 PFAS in humans are thought to be a function of strong tissue binding.⁶⁹ Differences between 322 human and rodent toxicokinetics have been proposed as factors affecting PFAS exposures. For 323 example, organic anion transporters (OATs),⁷⁰ organic anion transporting polypeptides (OATPs), 324 ⁷¹ and permeability have all been proposed as important determinants of human exposures. ⁴⁴ 325 326 However, PFAS bound in tissues are unavailable for elimination from the circulatory system through renal and biliary clearance. ³⁵ Notably, we found that $f_{unbound}$ correlates well with 327 elimination half-lives of PFAS in humans (Fig. S6). For example, the *f*_{unbound} of PFOS in human 328 tissues was 5.8-11.3 times lower than for PFOA, which corresponds to the difference in observed 329 elimination half-lives between the two PFAS (PFOA: 3.8 years, PFOS: 5.4 years).²³ 330

331 Previous rat and mouse studies reported liver PFOS concentrations that were ~20-fold higher than in lung and kidney tissue. 22,72 In contrast, our results suggest liver/kidney and 332 liver/lung concentration ratios of 0.53 and 1.5. Rodent studies using oral administration may not 333 334 reflect steady-state exposures in humans because PFAS pass the gastrointestinal barrier and 335 may accumulate in the liver after absorption from the gut. In this work, tissue/tissue concentration 336 ratios derived from experimental $K_{tissue/w}$ were compared to equivalent concentration ratios measured in human cohort studies. Interestingly, liver/tissue concentration ratios were similar 337 between binding experiments and in humans for the long-chain PFNA and PFOS ($\eta_{pfc} = 8$), 338 whereas PFHxS (η_{pfc} = 6) and PFOA (η_{pfc} = 7) were overestimated in the liver as compared to 339 340 kidney and lung tissue (Fig. S7). These comparisons indicate that for long-chain PFAS, tissue 341 binding is the major driver for elimination and tissue distribution in humans. Exposure to shortchain PFAS may be driven by complex kinetic processes like membrane transporters. 73,74 342 permeability, ⁴⁴ renal reabsorption, ⁷⁵ and hepatic recirculation. ⁷⁶ 343

Although the brain shows the highest sorption capacity for PFAS in our experiments, low concentrations of PFAS have been measured in human brains relative to liver, kidney, and lung

tissues (Fig. S7). ^{19,21} The lower relative tissue distribution of PFOS and PFOA to brain was also
observed in rats. ^{22,72} The limited distribution to brain *in vivo* is likely the result of the protective
effects of the blood-brain barrier⁷⁷⁻⁷⁹ because strong acids are typically restricted from crossing.
⁸⁰ Future studies are needed to assess to what extent tissue binding experiments reflect PFAS
exposures observed in other human tissues (e.g., spleen, mammary gland).

Implications for characterizing toxicokinetics of PFAS in humans. In this study, we 351 352 conducted binding experiments using five tissues from common animal model species (i.e., mouse and rat) and humans for 16 study PFAS to: (i) evaluate species and tissue-specific binding 353 and (ii) elucidate mechanistic understanding of factors affecting accumulation and elimination. 354 Results show only small differences in tissue binding across biological species. We find rat liver 355 provides a suitable surrogate tissue for predicting binding of PFAS in various human tissues 356 357 (kidney, brain, lung, heart) when used in combination with appropriate scaling factors. 358 Considering the large number and structural heterogeneity of PFAS, rat liver provides a useful 359 surrogate for screening and categorizing the internal distribution of PFAS in humans. To reduce 360 animal use, future studies should explore alternative surrogate materials, such as polymers 361 known to bind PFAS, to predict human tissue binding of these compounds. Data on $f_{unbound}$ and $K_{\text{tissue/w}}$ reported in this study can help parameterize physiologically-based toxicokinetic (PBTK) 362 models that are useful for quantifying the elimination and distribution of PFAS. However, it is 363 364 important to consider that other toxicokinetic processes, which may be more species-dependent than tissue binding, such as membrane transporter expression, could introduce inaccuracies 365 when applying the predicted *f*_{unbound} in human PBTK models. Still, human and modeling studies 366 both suggest tissue binding is a major factor driving PFAS exposure and elimination. ^{19,20} Future 367 studies should assess the variability in tissue $f_{unbound}$ among humans that result from genetic or 368 369 disease-related differences in tissue protein and lipid composition, such as lipid accumulation in fatty liver disease.⁸¹ The predictive framework presented in this study may be useful for assessing 370 371 potential accumulation of emerging PFAS with more complex structural and physicochemical

properties, which are increasingly detected in the environment. Understanding their distribution
 among human tissues will help inform regulations to control their production and release into the
 environment, reducing their uptake through drinking water and food.

375 Supporting Information

- 376 Information on PFAS chemicals tested in this study; LC-MS/MS quantification method; Unbound
- 377 fraction data; Pairwise comparisons of *f*_{unbound}; Statistical analyses; PLSR model results; Unbound
- 378 fractions compared to elimination half-lives and tissue distributions in humans.

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