

# Personalizing hemodialysis (HD) treatment in pediatric patients with end-stage renal disease (ESRD) – *application and integration of quantitative pharmacology with machine learning*

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# Pediatric Endstage Renal Disease (ESRD)

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- Mortality  $\geq 30x$   $\uparrow$  compared to healthy children
- Rare condition



# Hemodialysis (HD)

*= Mode of initial renal replacement therapy in ~ 50% of patients*

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## Adult targets:

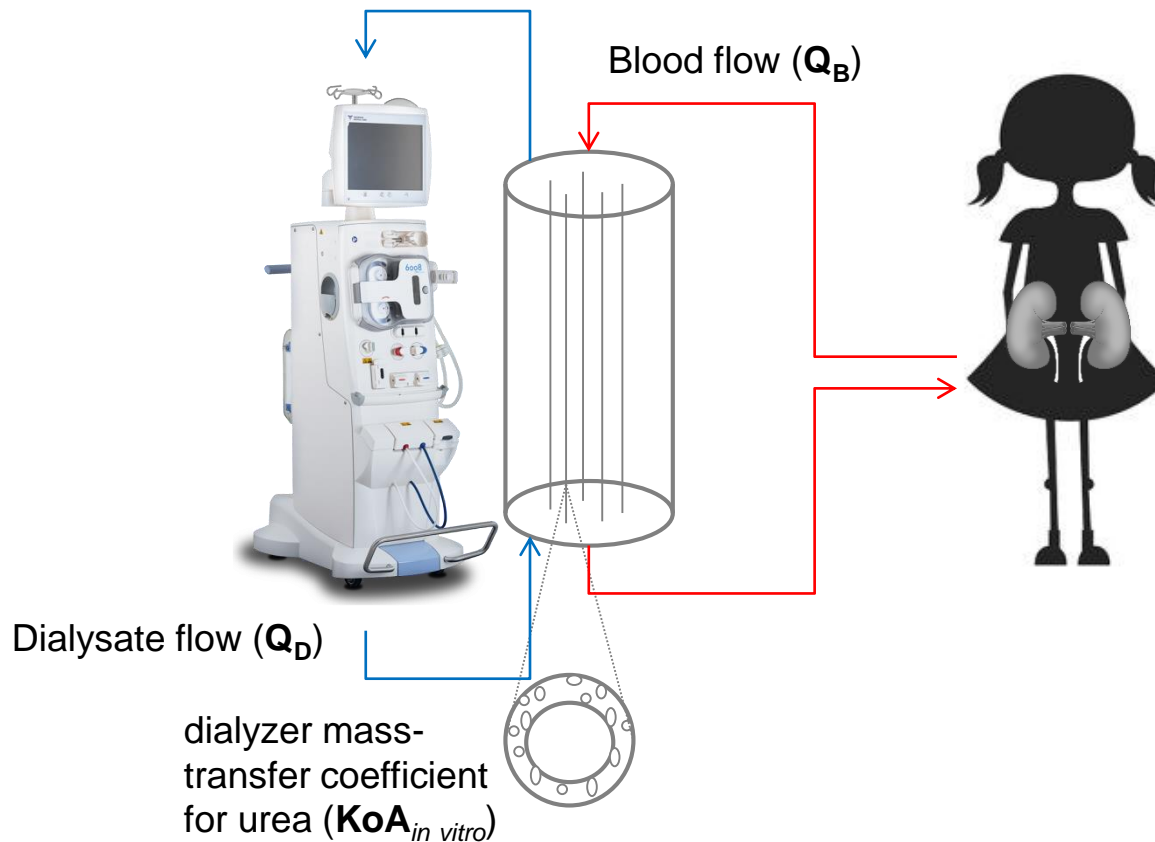
- Urea dialyzer clearance ( $K_D$ ): Weight-normalized  $Kt/V \geq 1.4$
- Fluid removal rate:  $< 10\text{-}12 \text{ ml/kg/h}$  (Ultrafiltration, **UFR**)



# Urea dialyzer clearance $K_D$

mechanistic prediction (mass balance) – needs in vivo correction

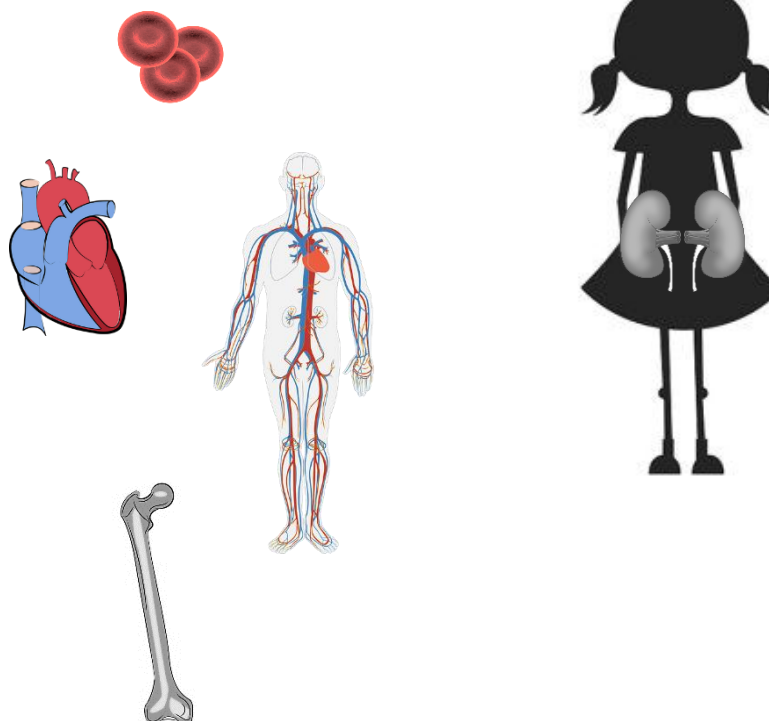
$$K_D = f(Q_D, Q_B, KoA_{in\ vitro})^1$$



# HD corrects only part of ESRD problems...

*Other disease-related problems require additional intervention strategies*

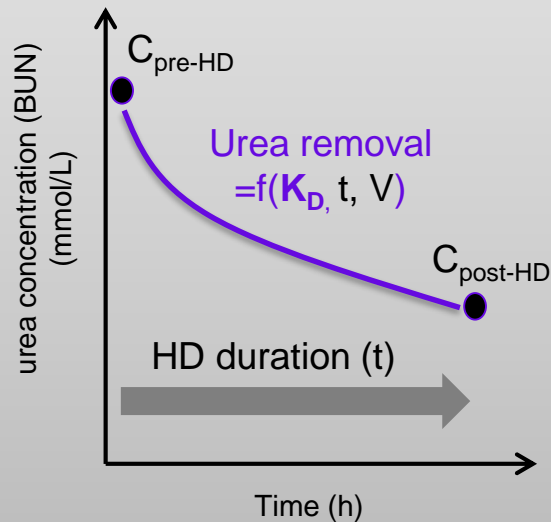
- Uremia
- Fluid overload
  
- Anemia
- Cardiovascular disease
- Hypertension
- Mineral and bone disorder
- Malnutrition
- ...



**Aim:** To develop scientific evidence for personalized HD treatment *in children*

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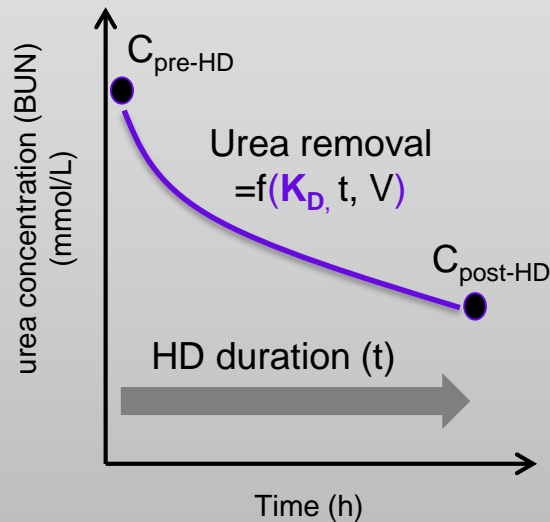
**Predict *in vivo* urea dialyzer clearance ( $K_D$ )**



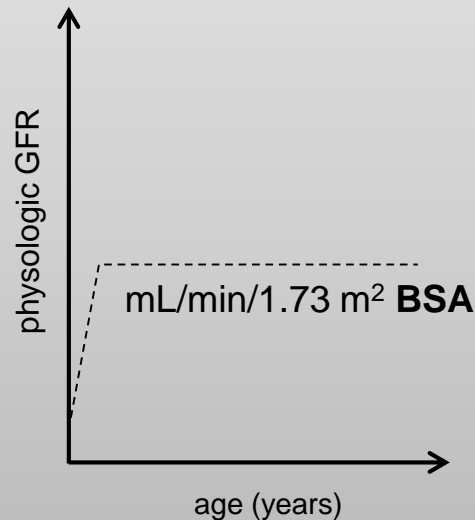


**Aim:** To develop scientific evidence for personalized HD treatment in children

**Predict *in vivo* urea dialyzer clearance ( $K_D$ )**

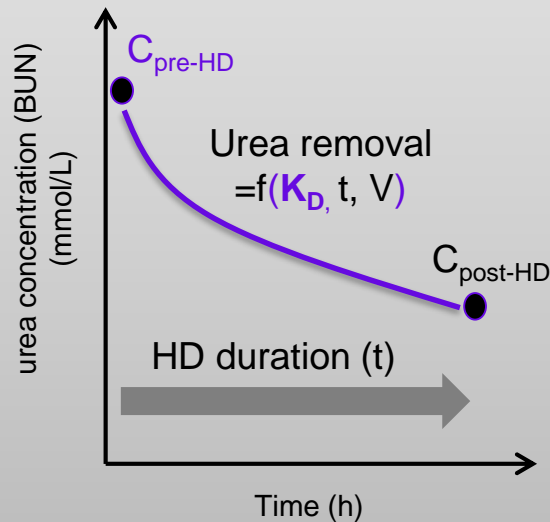


**Best HD dose ( $Kt/V$ ) and scaling ( $Kt/BSA$ )?**

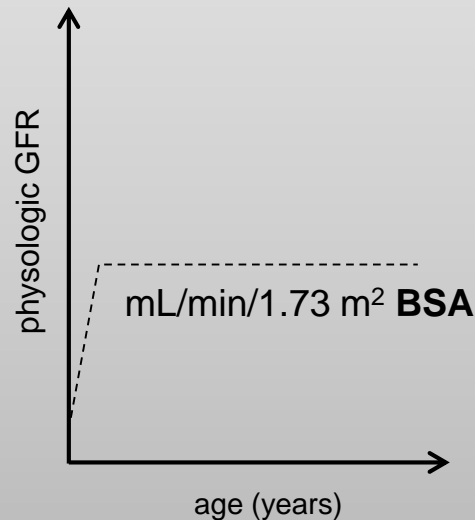


# Aim: To develop scientific evidence for personalized HD treatment in children

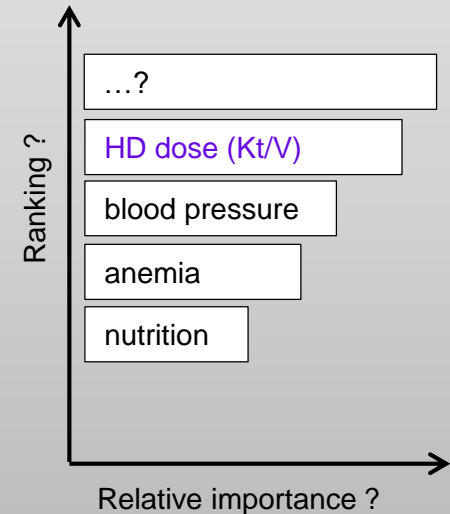
## Predict *in vivo* urea dialyzer clearance ( $K_D$ )



## Best HD dose ( $Kt/V$ ) and scaling ( $Kt/BSA$ )?



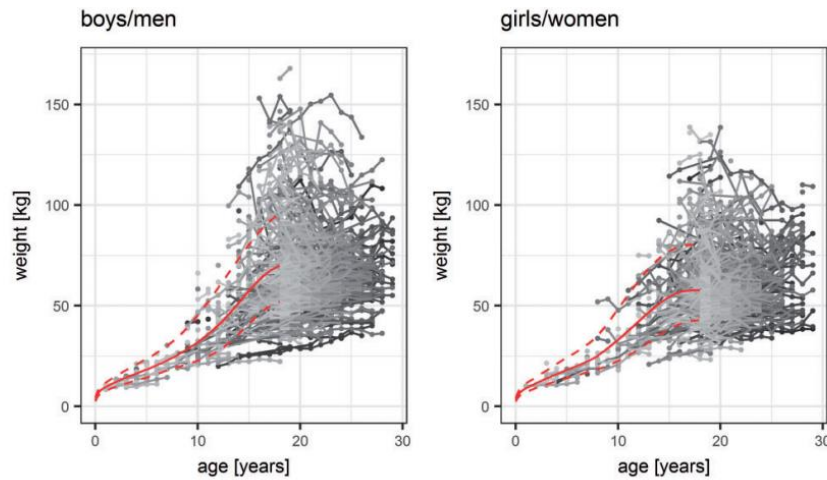
## Importance of HD dose ( $Kt/V$ ) compared to other factors?



# Data used for retrospective analysis

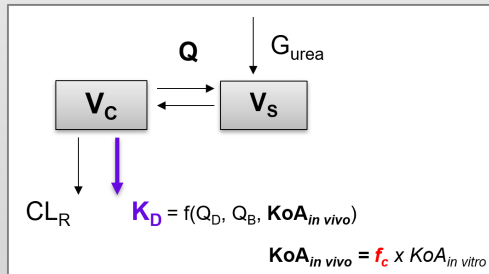
## Cohort of 1852 patients

- ✓ on chronic HD since childhood (<19 years)
- ✓ HD 3x/week
- ✓ <30 years (2004-2016)

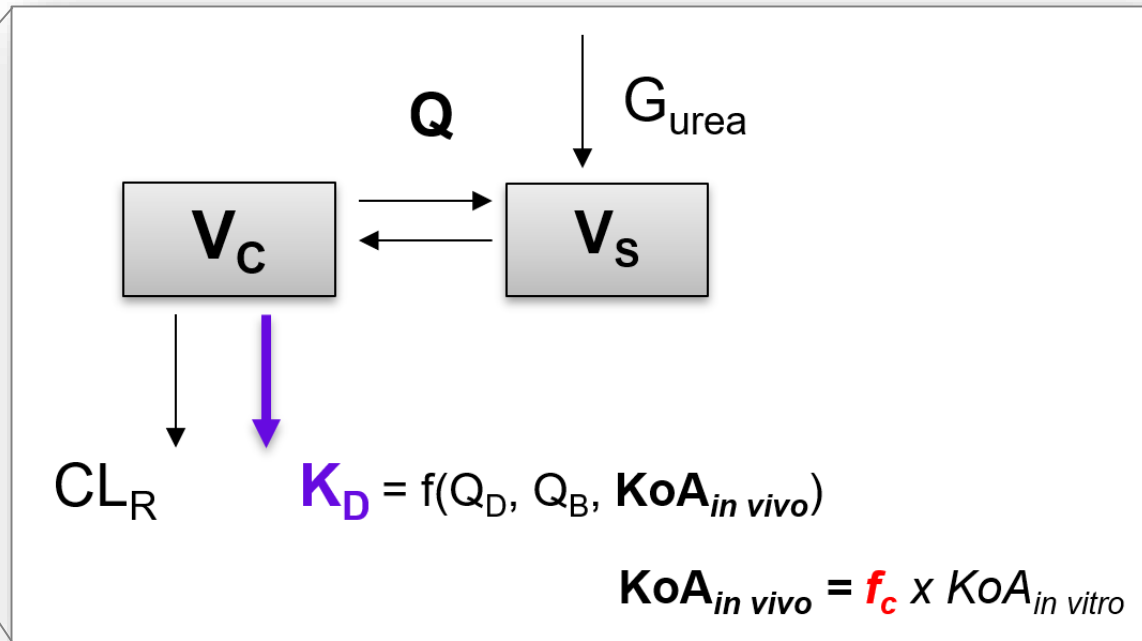


# Methods

## Pharmacometric (PMX) modeling of $K_D$

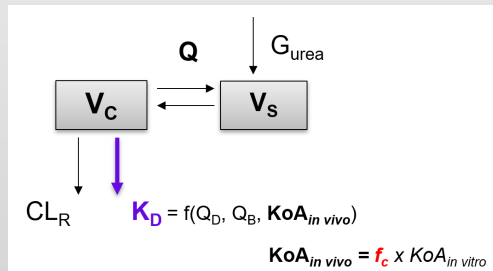


- Physiologic scaling of distribution parameters (perfusion-limited)
- Mechanistic  $K_D$  prediction from  $KoA_{in vitro}$
- Covariate analysis: factors associated with *in vivo* correction factor  $f_c$



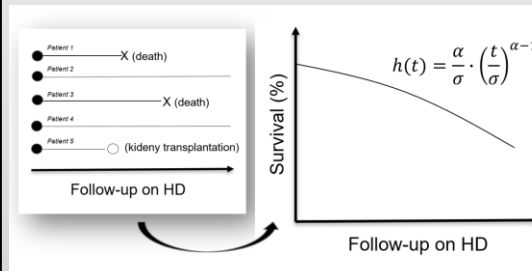
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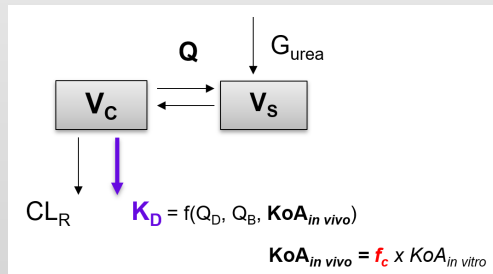
## Parametric time to event (TTE) survival modeling



- Weibull accelerated failure time (AFT) model
- mean  $Kt/V$  versus  $Kt/BSA$  as predictors of log hazard

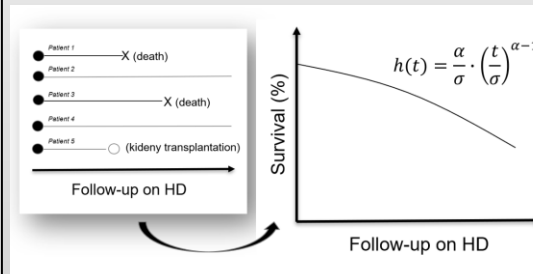
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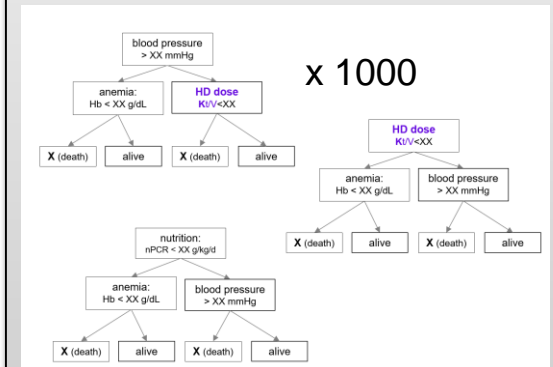
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## Parametric time to event (TTE) survival modeling



- Weibull accelerated failure time (AFT) model
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## Machine learning (ML) 5-year mortality



- Random forest
- **HD dose ( $Kt/V$ )** and > 100 other potential predictor variables (features)

# Results

## *Included data*

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N=1852 patients (53903 HD treatments)



**Pharmacometric (PMX)  
modeling of  $K_D$**

N=1691 patients

(923+768 for model  
development+evaluation)

**Parametric time to event  
(TTE) survival modeling**

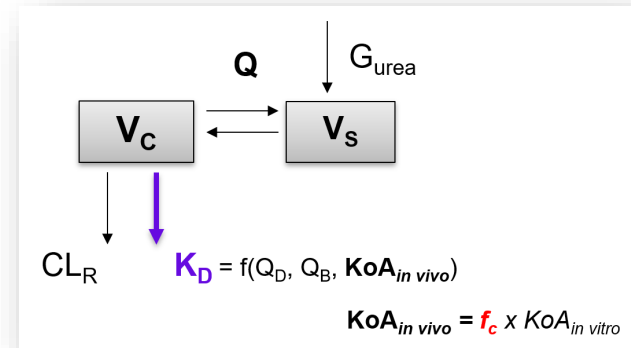
N=1493 patients

**Machine learning (ML)  
5-year mortality**

N=363 patients

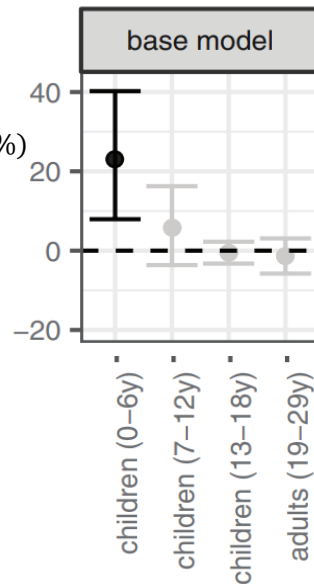
# PMX modeling of $K_D$

**Base model:**  $C_{post-HD}$  underpredicted in children <6 years



$f_c = 1.2$  (RSE: 1.6%)

**Mean prediction error (%)**  
 $= \frac{C_{IPRED} - C_{OBS}}{C_{OBS}} \cdot 100$

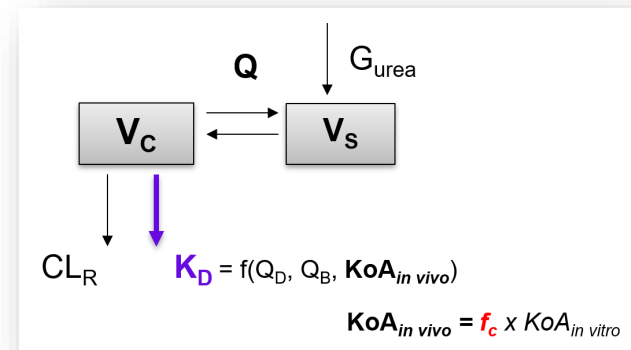
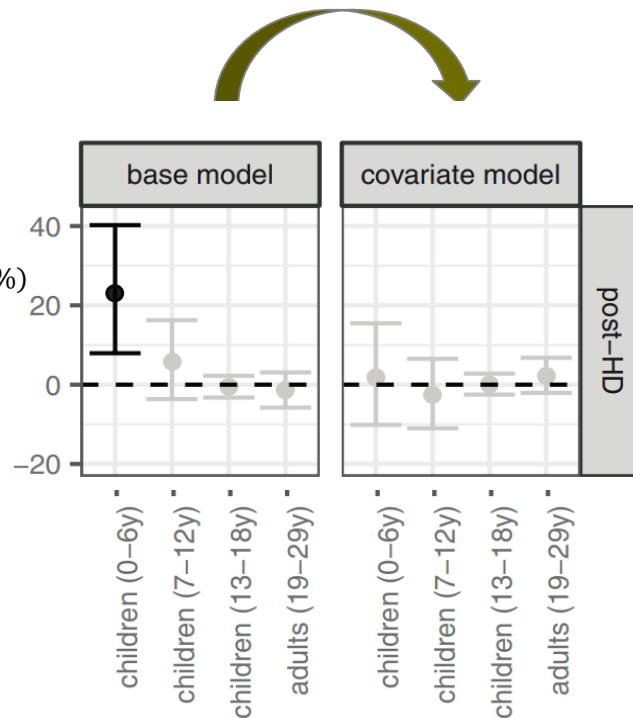




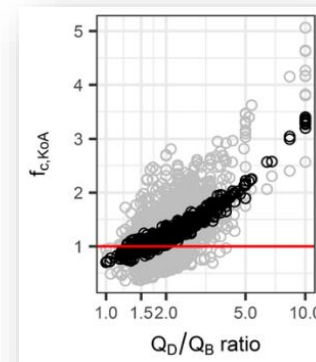
# PMX modeling of $K_D$

**Covariate model:** accounting for high  $Q_D/Q_B$  corrects bias in children <6 years

$$\text{Mean prediction error (\%)} = \frac{C_{IPRED} - C_{OBS}}{C_{OBS}} \cdot 100$$



$$f_c = f(Q_D/Q_B \text{ ratio})$$



Population predictions  
 Individual predictions

# PMX modeling of $K_D$

## **Covariate model:** other factors for personalized in vivo HD clearance prediction

- ✓  $Q_D/Q_B$  ratio
- ✓ Type of filter (low-/high-flux)
- ✓ Predicted true  $Q_B$   
(lower than nominal  $Q_B > 200$  mL/min)

### Typical adult prescription

Prescription	
Blood flow ( $Q_b$ )	500 mL/min
Dialysate flow ( $Q_d$ )	800 mL/min
filter mass-transfer-area coefficient for urea (KoA in vitro)	800 mL/min
Low-flux filter use	0   1=yes, 0=high-flux
Calculated values	
Calculated in vitro urea dialyzer clearance ( $K_D$ ) without correction*	274 mL/min
Calculated true $Q_b$ **	457 mL/min
Calculated $Q_d/Q_b$ ratio	1.60
Calculated KoA correction factor for $Q_d/Q_b$ ratio and filter flux	0.66
<b>Calculated corrected in vivo urea dialyzer clearance (<math>K_D</math>)***:</b>	<b>198 mL/min =72.1% of in vitro <math>K_D</math></b>

### Pediatric prescription

100 mL/min
500 mL/min
300 mL/min
0   1=yes, 0=high-flux
44 mL/min
100 mL/min
5.00
2.06
<b>68 mL/min =156% of in vitro <math>K_D</math></b>

# PMX modeling of $K_D$

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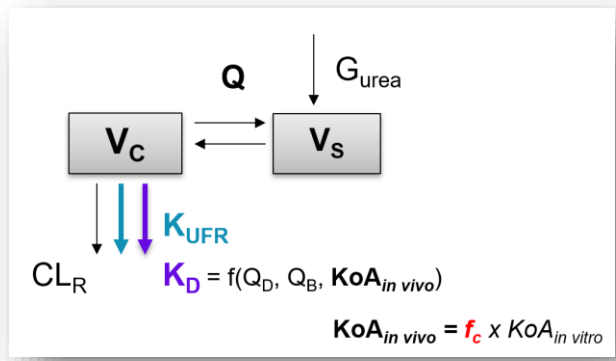
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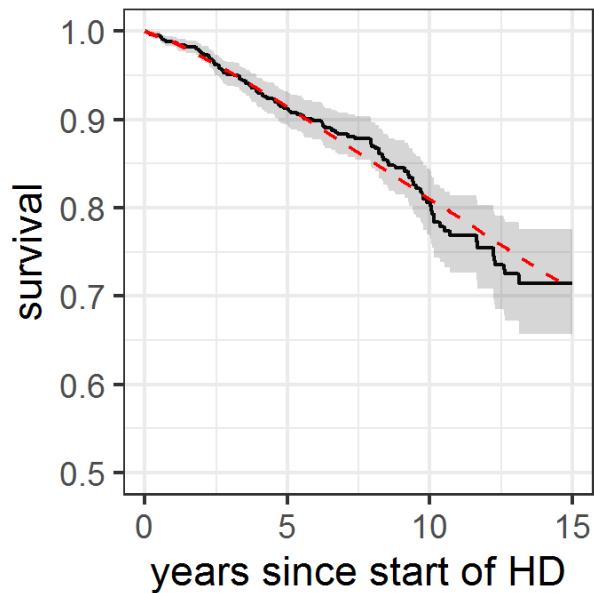
- ✓ **Ultrafiltration rate ( $K_{UFR}$ ):** adds as convective clearance to diffusive  $K_D$



$$CL_{tot} = CL_R + K_D + K_{UFR}$$

# Parametric TTE modeling

*Weibull model predicts baseline hazard well*



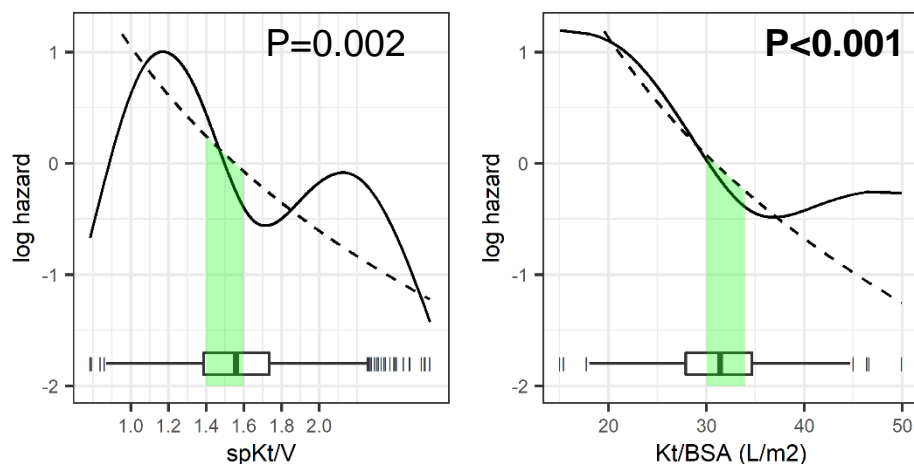
--- Weibull model predictions (unadjusted)

- Scale  $\sigma = 35.4$  years  
(time when predicted survival = 40%)
- Shape  $\alpha = 1.23$   
(indicating with  $\alpha > 1$  increasing hazard of death over time)

— Kaplan Meier curve (95% CI)

# Parametric TTE modeling

*BSA-based HD dose (Kt/BSA) better predictor of survival than weight-based Kt/V*

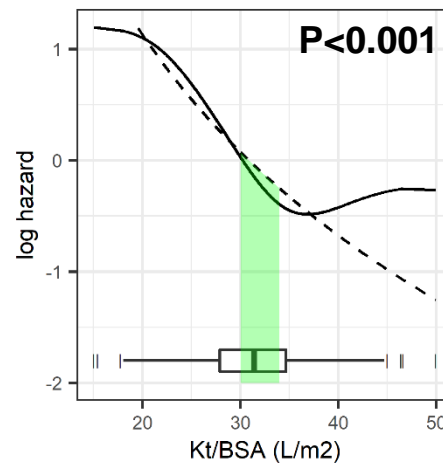
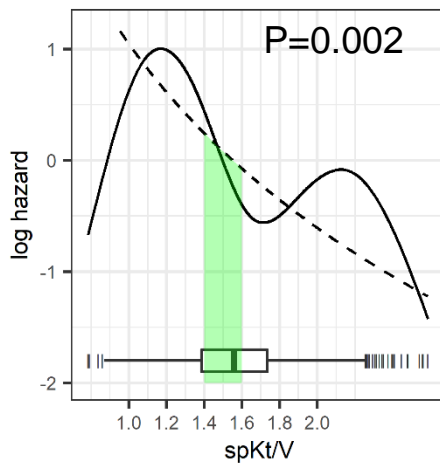


- - - Weibull model prediction (log-linear relationship with log hazard)

\_\_\_ prediction from flexible non-linear spline model

# Parametric TTE modeling

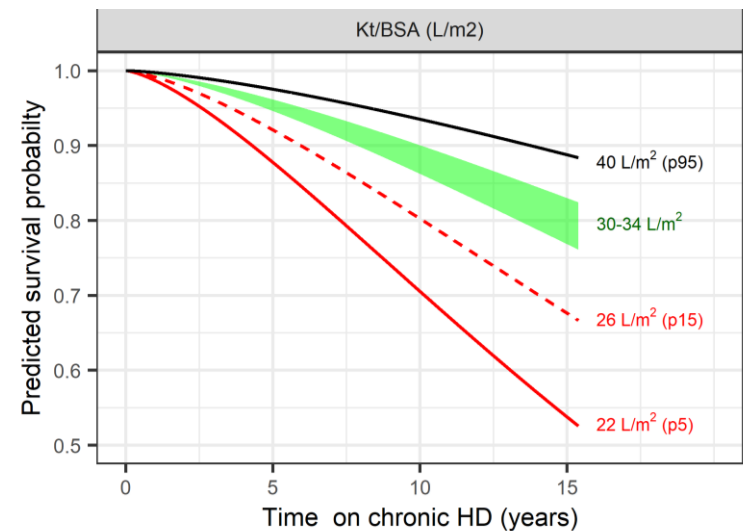
BSA-based HD dose (Kt/BSA) better predictor of survival than weight-based Kt/V



**Kt/BSA >30 L/m<sup>2</sup>** corresponds to

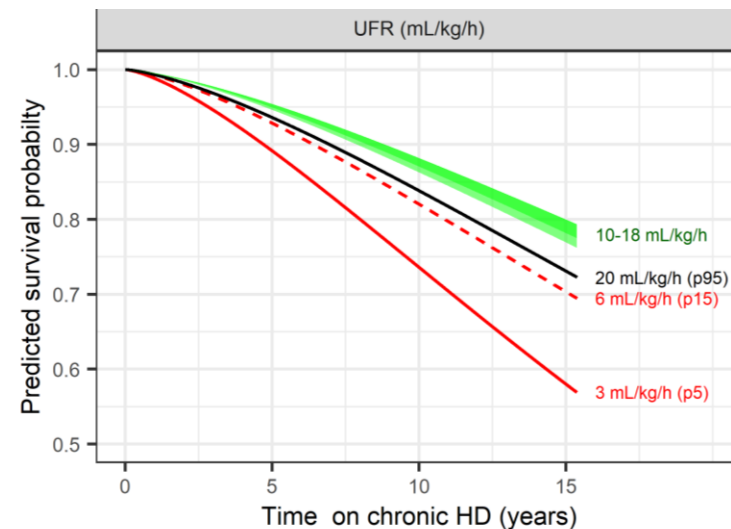
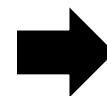
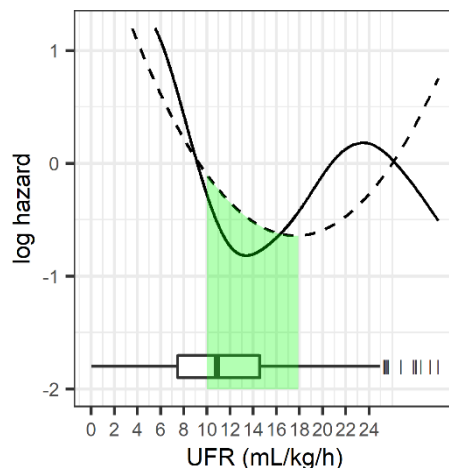
- Kt/V >1.4 (>12 years / adults)
- Kt/V >1.6 (<12 years)

- - - Weibull model prediction (log-linear relationship with log hazard)  
 \_\_\_\_ prediction from flexible non-linear spline model



# Parametric TTE modeling

## *Ultrafiltration (UFR) associated with survival in U-shaped relationship*



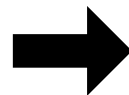
- - - Weibull model prediction (quadratic relationship with log hazard)

\_\_\_ prediction from flexible non-linear spline model

# Machine learning (Random forest)

**12 predictors related to nutrition, inflammation, anemia and HD dose (Kt/V, UFR)**

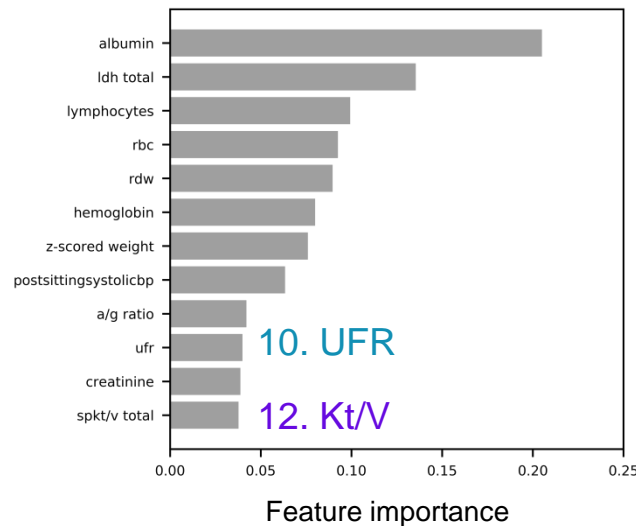
105 potential predictor variables («features»)



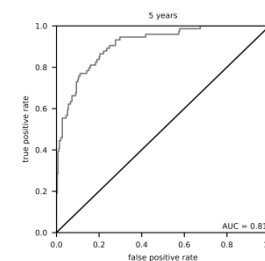
Final ML model: n=12 features retained



- Demographics
- HD treatment
- Laboratory measurements (monthly)



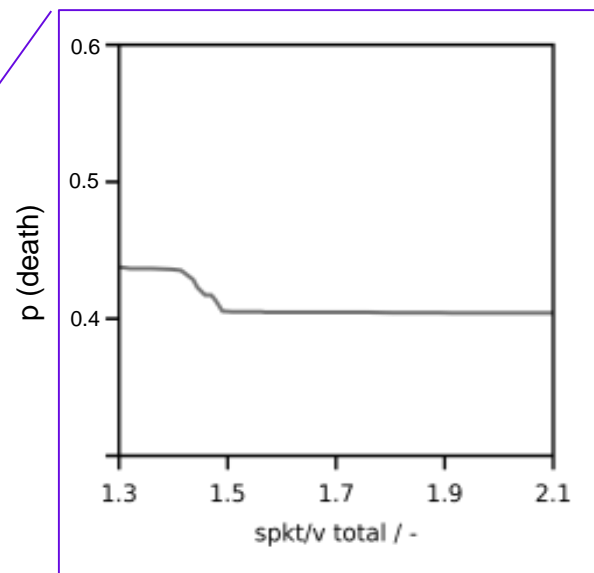
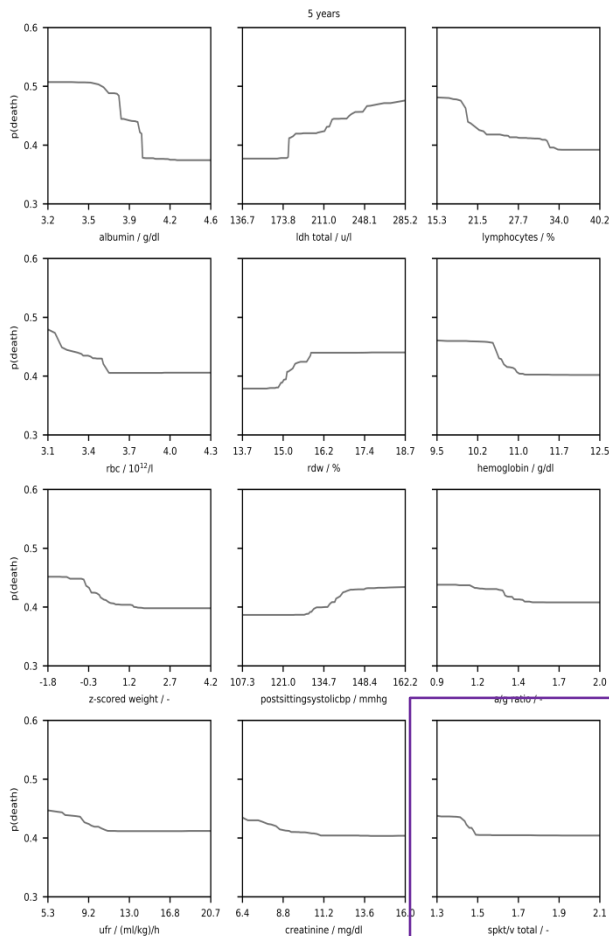
AUC<sub>ROC</sub>: 0.81





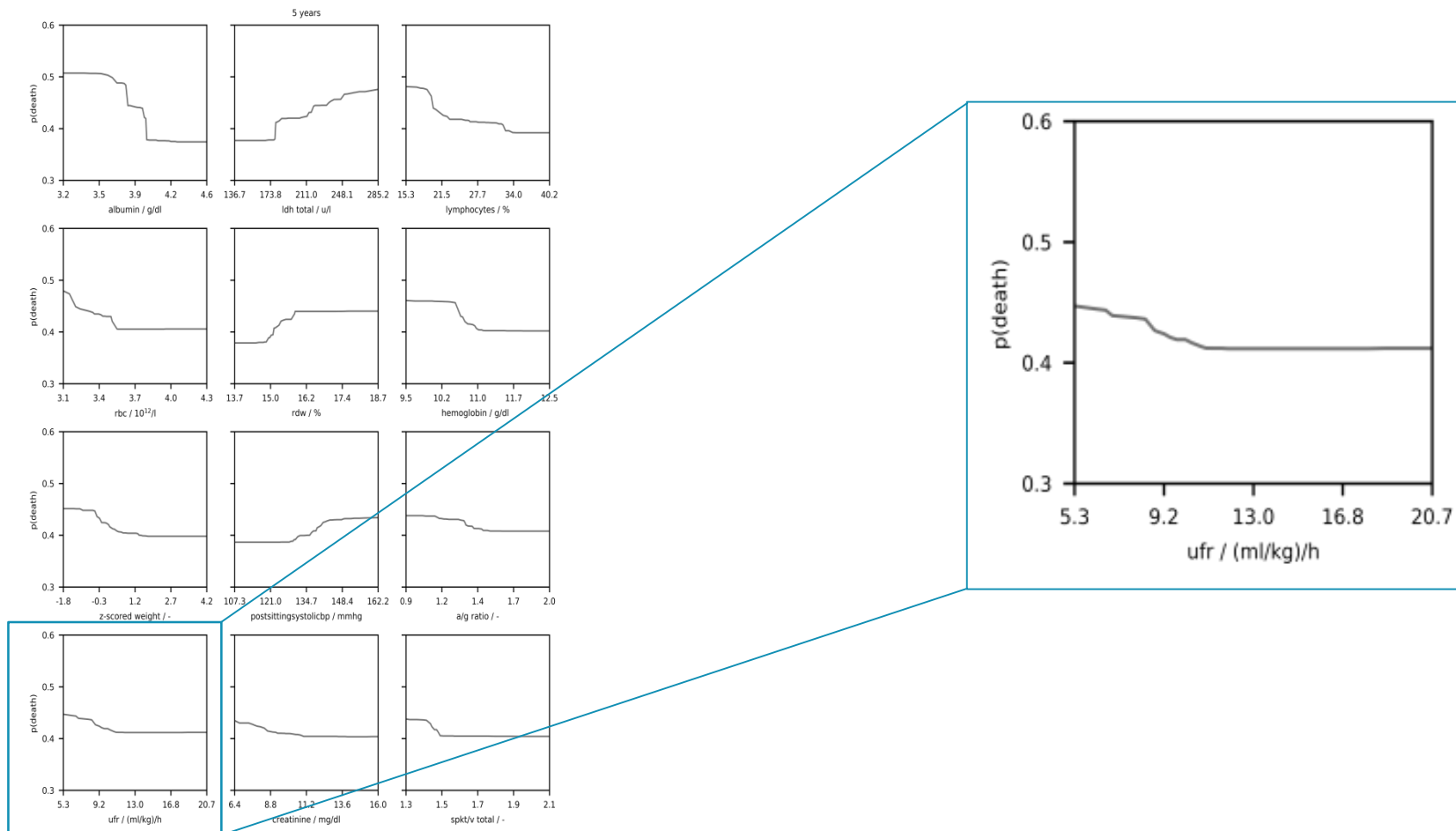
# Machine learning (Random forest)

**Partial dependence plots: Increased mortality with low  $Kt/V < 1.5$**



# Machine learning (Random forest)

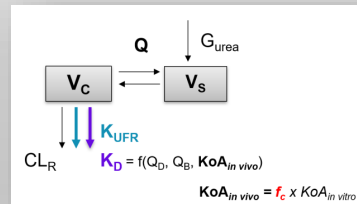
**Partial dependence plots: Increased mortality with low UFR < 10 mL/kg/h**



**Conclusion:** Quantitative pharmacology and ML approaches can help to personalize HD treatment in children

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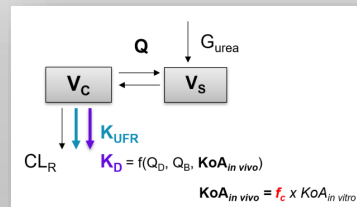
**PMX:** Scaling/predicting urea dialyzer clearance ( $K_D$ ) from adult to pediatric HD patients



**TTE/ML:** Intense HD prescription in children needed for best long-term survival

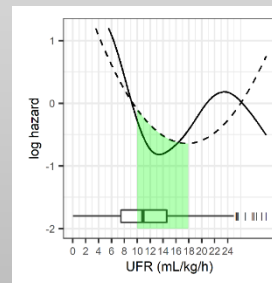
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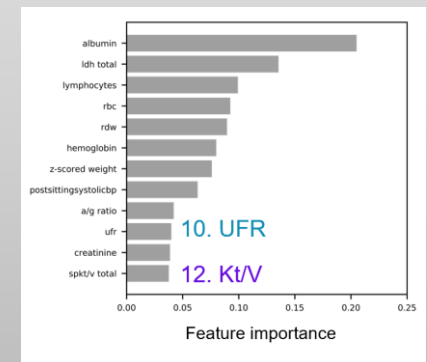


**TTE/ML:** Intense HD prescription in children needed for best long-term survival

- $Kt/BSA > Kt/V$  (alternatively: age-dependent  $Kt/V$ )
- UFR: U-shaped relationship (increased mortality  $<10$  and  $>18$  mL/kg/h)



- Importance of other disease-related factors besides HD dose ( $Kt/V$ ) / UFR



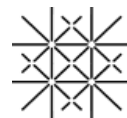
# Thank you for your attention!



Prof. Marc Pfister



Dr. Andrew Atkinson



Universität  
Basel

Forschungsfonds



Prof. Olivera Marsenic  
Coloures



Lucile Packard  
Children's Hospital  
Stanford



Prof. Julia Vogt



Georgi Tancev



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