

# Maternal-Fetal PBPK Modeling of Antipsychotic Drugs as Case Study

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# Disclaimer

This presentation is the opinion of the author only and does not represent FDA policy and regulation

# Perinatal Health and Regulatory Need

- Need for computational alternatives to support the evaluation of safety and efficacy of therapeutic agents for patients in perinatal life-stages who are frequently excluded from clinical trials
- Utilize life-stage PBPK models as tools to evaluate and optimize treatments specifically in the perinatal life-stages
- The confidence in the use of such tools in regulatory decision-making and pose challenges in conducting timely and thorough review of submissions

# Project Goal and Purpose

- To develop a PBPK model for different perinatal life stages (fetuses, neonates, infants, young children, pregnant and lactating women)
- To serve as an in-house tool allowing for flexibility to conduct any investigative analysis, and to better inform dosing for antipsychotics and COVID-19 treatments in these data-sparse populations



# Strategy for Model Development



**Curate data:** Gather available information (existing databases, body of literature, and FDA in-house data) for parameterization of perinatal life-stage PBPK models

## **Design model:**

- Maternal Component
- Fetal Component
- Population Modeling
- Other Perinatal Life-stages including Pediatrics

**Code model component:** Code the model using an open-source language based on the determined model structure and parameters. Conduct sensitivity analysis to confirm model dimensionality

**Test model component:** Verify the model's predictive capability using case studies

**Document model and scope gap/needs:** Document model code including assumptions and decisions. Identify gaps in data and modeling needs to guide future research to facilitate regulatory application of PBPK modeling tools in perinatal periods

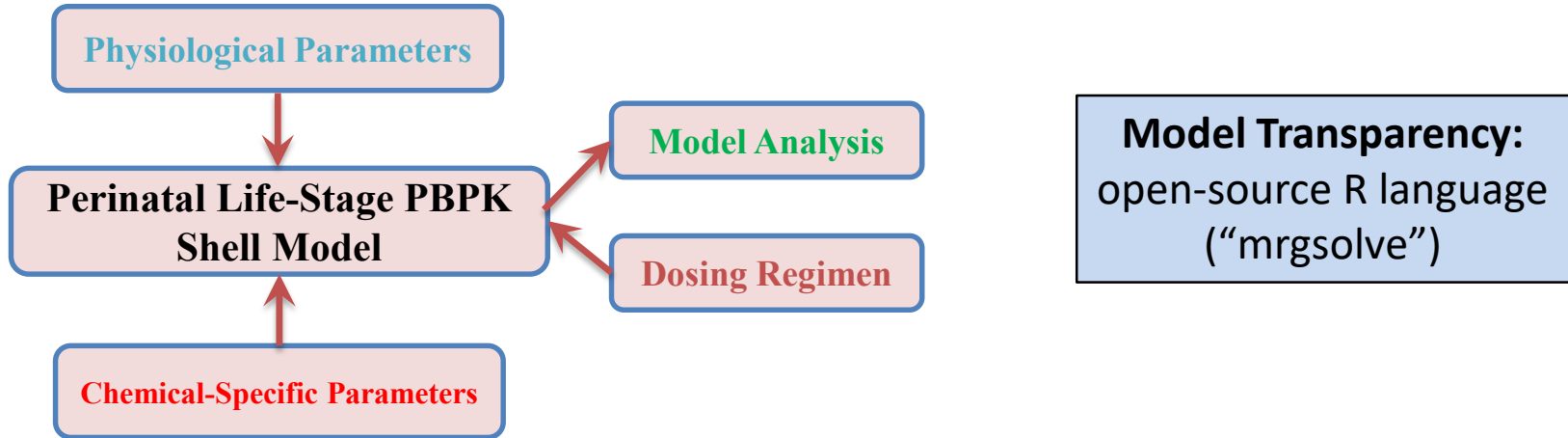
# Repository for Physiological Changes

## Equations for Maternal Fetal Physiological Changes

Parameters	Abduljalil et al. 2012 & 2019	Dallmann et al. 2017	Kapraun et al. 2019
Total Body Fat (kg)	$17.14+0.1305*GA+0.0008*GA^2$ (Eq 9)	$20.2+0.125*FA$ (Eq 15)	$17.067+0.14937*GA$ (Eq 3)
Cardiac Output (L/h)	$301+5.916*GA-0.088*GA^2$ (Eq 14)	$6.09\exp(-\exp(-0.352\log FA + 1.36)) + 5.14$ (Eq 19) L/min)	$301.78+3.2512*GA+0.15947*GA^2-0.0047059*GA^3$ (Eq 13)
Plasma Volume (L)	$2.50-0.0223*GA+0.0042*GA^2-0.00007*GA^3$ (Eq 15)	$2.36+0.0000841*FA+0.00395*FA^2-0.0000817*FA^3$ (Eq 29)	$2.4958+1.2406/(1+\exp(-0.31338*(GA-17.813)))$ (Eq 5)
Hematocrit (%)	$39.1-0.0544*GA-0.0021*GA^2$ (Eq 17)	$40.1+0.0299*FA-0.0180*FA^2+0.000401*FA^3$ (Eq 21)	$39.192-0.10562*GA-(7.1045*10^{-4})*GA^2$ (Eq 24)
Albumin (g/L)	$45.8-0.1775*GA-0.0033*GA^2$ (Eq 19)	$14.7*\exp(-0.0454*FA)+31.7$ (Eq 9)	NA
Alpha-1-Acid Glycoprotein AAG (g/L)	$0.74-0.0088*GA+0.0001*GA^2$ (Eq 20)	$0.0000768*FA^2-0.00573*FA+0.701$ (Eq 11)	NA
Glomerular Filtration Rate	$114+3.2367*GA-0.0572*GA^2$ (mL/min, Eq 27)	$0.113+8.73*FA-0.716*FA^2+0.0245*FA^3-0.000294*FA^4$ (FA<=14) $0.15$ (FA>=14) (L/min, Eq 27)	$(17+(12.5-17)/40*GA)/100$ (unitless, Eq 16)
Placenta Volume	$0-0.716*GA+0.9149*GA^2-0.0122*GA^3$ (mL, Eq 43)	$0.937*\exp(-6.10*\exp(-0.0813*FA))-0.00211$ (L, Eq 48)	$0-1.7646*GA+0.91775*GA^2-0.011543*GA^3$ (mL, Eq 8)
Amniotic Fluid Volume	$1.9648*GA-1.2056*GA^2+0.2064*GA^3-0.0061*GA^4-0.00005*GA^5$ (mL, Eq 44)	$\exp(-10+1.22*GA-0.0653*GA^2+0.00169*GA^3-0.0000169*GA^4)$ (L, Eq 33)	$822.34/(1+\exp(-0.26988*(GA-20.150)))$ (mL, Eq 9)
Fetal Tissue Volume	$0.01*\exp((0.955/0.0702)*(1-\exp(-0.0702*GA)))$ (mL, Eq 42)	$5.90/(1+\exp(-4.45*\log FA + 15.8))$ (L, Eq 39)	NA

Abduljalil et al. 2012 (PMID: 22515555); Abduljalil et al. 2019 (PMID: 29987449); Dallmann et al. 2017 (PMID: 28401479); Kapraun et al. 2019 (PMID: 31048866); Zhang et al. 2017 (PMID: 28588050)

# Code the Pregnancy PBPK Model in R



## Shell Model (generic base model)

```

104 //change of fraction unbound in plasma
105 double fup = fup_0*(1+kaff*AAO_D)/(1+kaff*AAO);
106
107 //liver clearance
108 double CL_Li = ((VmaxH/km)*MPPLG*Vli*1000*60*1e-6)/fumic*Activity_CYP3A4;
109
110 [ODE]
111 //calculation of tissue drug concentrations (mg/L)
112 double cblood = BLOOD/vb;
113 double cadipose = ADIPOSE/vad;
114 double ckidney = KIDNEY/vki;
115 //double cgutlumen = GUTLUMEN/vgulumen;
116 double cgut = GUT/vguwll;
117 double crichly = RICHLY/vvrp;
118 double cpoorly = POORLY/vpp;
119 double cliver = LIVER/vli;
120 double cpfpu = PFU/(vpfu+PF);
121
122 //capillary blood concentration
123 double cvad = cadipose/kpad*BP;
124 double cvgu = cgrt/Kppu*BP;
125 double cvki = ckidney/kpki*BP;
126 double cvrp = crichly/kprp*BP;
127 double cvpp = cpoorly/kppp*BP;
128 double cvli = cliver/kli*BP;
129 double cvpfu = cpfpu/kppfu*BP;
130 double cven = (qad*cvad + qki*cvki + qrp*cvrp + qpp*cvpp + qli*cvli + qpfu*cvpfu)/QC;
131
132 //ODES
133 dxdt_GUTLUMEN = -(ka*ki)*GUTLUMEN;
134 dxdt_GUT = ka*GUTLUMEN + qguwall*(cblood - Cvgu);
135 dxdt_ADIPOSE = qad*(cblood - cvad);
136 dxdt_KIDNEY = qki*(cblood - cvki) - CL_Ki*(Fup*cvki);
137 dxdt_RICHLY = qrp*(cblood - cvrp);
138 dxdt_POORLY = qpp*(cblood - cvpp);
139 dxdt_LIVER = qguwall*(cvgu + (q11*qguwall)*cblood - q11*cvli - CL_Li*(Fup*cvli));
140 dxdt_PFU = qpfu*(cblood - cvpfu);
141 dxdt_CL = CL_Ki*(Fup*cvki) + CL_Li*(Fup*cvli);
142 dxdt_BLOOD = qc*(cven-cblood);
143 dxdt_ABS = ka*GUTLUMEN;
144
145 double massbalance = BLOOD*ADIPOSE+KIDNEY*GUT+RICHLY+POORLY+LIVER+PFU*CL_ABS;
146 double SumQ = (Qad + Q11 + Qk1 + Qrp + Qpp + Qc);
147 double SumV = (Vb1 + Vad + vguwall + Vli + vki + vpfu + vrp + vpp)/Bw;
  
```



## 1. Read Shell Model + Source the Equations

```

8 library(tibble)
9 library(tidyverse)
10 library(magrittr)
11
12 setwd("c:/pPBPK/shell model/darunavir")
13 source("../equations for physiological parameters.R")
14 darunavir_oral_600 <- read_excel("darunavir_oral_600.xlsx") %>% as.data.frame()
15 mod <- mread_cache("Darunavir")
16 param(mod)
17
  
```

## 2. Generate Physiological Parameters

```

27 pPBPK_para <- tibble(id = rep(1:N, each = end/delta+1),
28   GA = rep(seq(GA_D, GA_D+(N-1)*10, 10), each = end/delta+1),
29   TIMEH = rep(TIMEH_seq, times = N)) %>%
30   mutate(GABW = GA * (timeH/7.24)) %>% # AGE2
31   mutate(BW = BW_A(GABW)) %>%
32   mutate(QCC = QCC_D(GA_to_FA(GABW))/BW^0.75) %>%
33   mutate(vbrC = vbrC_K, vguwallC = vguwallC_K, vliC = vliC_K, vskC = vskC_D, vteC = vteC_K) %>%
34   mutate(vbrD = vbrD(GA_to_FA(GABW))/BW, vbrC = vbrC(GA_to_FA(GABW))/BW, vbrE = vbrE_D(GA_to_FA(GABW))/BW,
35     vbrEC = vbrE_D(GA_to_FA(GABW))/BW, vbrEC = vbrE_D(GA_to_FA(GABW))/BW, vteC = vteC_D(GA_to_FA(GABW))/BW,
36     vteC = vteC_D(GA_to_FA(GABW))/BW) %>%
37   mutate(qadC = qadC_D, qkC = qkC_D) %>%
38   mutate(qbrC = qbrC_D(GA_to_FA(GABW)), qguwallC = qguwallC_K(GABW), q11C = q11C_K(GABW), q11C = q11C_K(GABW),
39     qskC = qskC_D(GA_to_FA(GABW))) %>%
40   mutate(qk1C = qk1_D(GA_to_FA(GABW)), qk1C = qk1_D(GA_to_FA(GABW)))/(QCC*BW), q11C = q11_D(GA_to_FA(GABW)))/(QCC*BW),
41     qbrC = (qbr_D(GA_to_FA(GABW)))/vbrC/(QCC*BW), qbrC = qbr_D(GA_to_FA(GABW))/vbrC/(QCC*BW),
42     qbrC = qbr_D(GA_to_FA(GABW))/vbrC/(QCC*BW) %>%
43   mutate(ActiVity_CYP3A4 = ActiVity_CYP3A4_A(GABW)) %>%
44   mutate(AAG = AAG_D(GABW)) %>%
45   mutate(PF = rep(1, N*(end/delta+1)))
  
```

## 3. Set the Dosing Regimen + Run the Simulation

```

52 pPBPK_para %>%
53   mutate(EVID = if_else((timeH %>% i) == 0 & timeH <-add1*i, true = 1, false = 0)) %>%
54   mutate(AMT = case_when(EVID == 1 ~ 600)) %>%
55   mutate(CMT = if_else(EVID == 1, true = 1, false = 2)) %>%
56   mutate(TIME = timeH)
57
58 simulation <- mod %>% mrgsim_d(data=pPBPK_para) %>% as.data.frame() %>% slice(3:N)
59
  
```

# Pregnancy PBPK Model

## Abduljalil et al. 2012

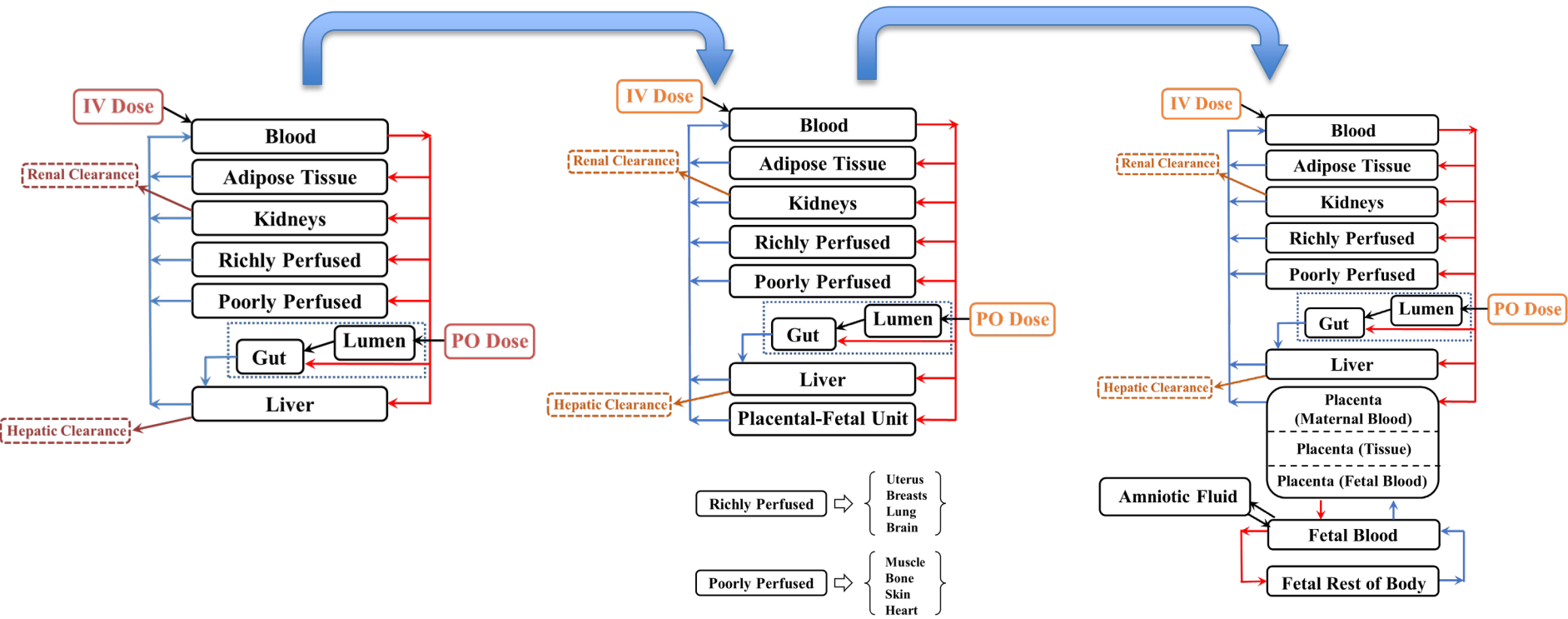
$$\begin{aligned}
 BW &= 61.1 + 0.2409 \times GA + 0.0038 \times GA^2 \\
 CO &= 301 + 5.916 \times GA - 0.088 \times GA^2 \\
 &\dots
 \end{aligned}$$

## Dallmann et al. 2017

$$\begin{aligned}
 CO &= 6.09 \times \exp(-\exp(-0.352 \times \lg(FA) + 1.36)) + 5.14 \\
 HCT &= 40.1 + 0.0299 \times FA - 0.0180 \times FA^2 + 0.000401 \times FA^3 \\
 &\dots
 \end{aligned}$$

## Customized Equations

$$\begin{aligned}
 BW &= 61.103 - 0.010614 \times GA + 0.029161 \times GA^2 - (5.0203 \times 10^{-4}) \times GA^3 \\
 CO &= 301.78 + 3.2512 \times GA + 0.15947 \times GA^2 - 0.0047059 \times GA^3 \\
 &\dots
 \end{aligned}$$



Non-Pregnant Adult Model

Pregnancy Model

Maternal-Fetal Model



## Application of the Model to Antipsychotic Drugs

- Antipsychotics are widely used during pregnancy and sometimes for the pre-existing condition
- Antipsychotics are often off-label use for pregnant women
- Antipsychotics are being prescribed increasingly during pregnancy and the postpartum period
- The dose-dependent drug effects and toxicity of antipsychotics on fetuses are still lacking
- Concerns and debates for birth defects using of antipsychotics during the first trimester
  - **Aripiprazole**: High Protein Binding; Multiple Metabolic Enzymes (with more than one metabolites)
  - **Risperidone**: High Protein Binding; Multiple Metabolic Enzymes
  - **Clozapine**: High Protein Binding; Multiple Metabolic Enzymes

# Application of the Model to Antipsychotic Drugs

Drugs	Human fup	logP	MW	pKa	Kp	Metabolism	BDDCS	ECCS	Available PBPK	Available pregnancy PBPK	Available pregnancy PK data
Aripiprazole	0.01	5.21	448.4	7.6	All Tissue to Plasma Partition	CYP3A4 CYP2D6	2	2	Yes (Vieira et al. 2014)	No	Yes
Risperidone	0.1	4.07	410.5	8.2	Coefficients were Calculated using Rodgers and Rowland Method in Simcyp	CYP2D6 (major) CYP3A4 (minor)	1	2	Yes (Kneller et al. 2020)	No	Yes
Clozapine	0.05	3.21	326.8	7.5		CYP1A2 CYP2D6 CYP3A4	2	2	Yes (Wong et al. 2019)	No	Yes

## Data Available for Non-Pregnancy Model

### Available adult PBPK models

**Aripiprazole:** Vieira et al. 2014 (PMID: 24556783)

**Risperidone:** Kneller et al. 2020 (PMID: 31359271)

**Clozapine:** Wong et al. 2019 (PMID: 30676661)

## Data Available for Pregnancy Model

### Available PK data for non-pregnant adults and pregnancy

**Aripiprazole:** Boulton et al. 2018 (PMID: 18563956); Westin et al. 2018 (PMID: 28643331)

**Risperidone:** Novalbos et al. 2010 (PMID: 20814331); Westin et al. 2018 (PMID: 28643331)

**Clozapine:** Hägg et al. 1999 (PMID: 10379638); Westin et al. 2018 (PMID: 28643331)

Values of physicochemical parameters are from DrugBank and PubChem

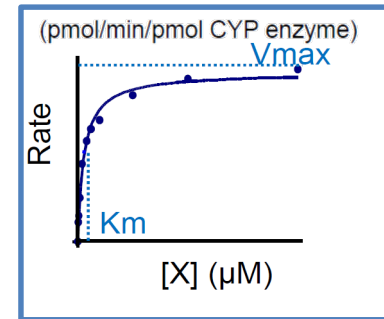
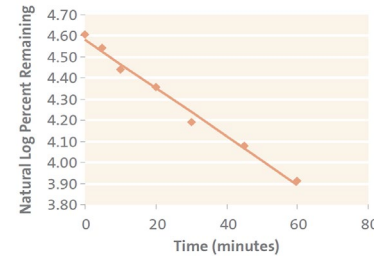
# in vitro to in vivo Extrapolation (IVIVE)

## Recombinant CYPs



### In Vitro $CL_{int}$

$\mu\text{L} / \text{min} / \text{per functional unit of system}$



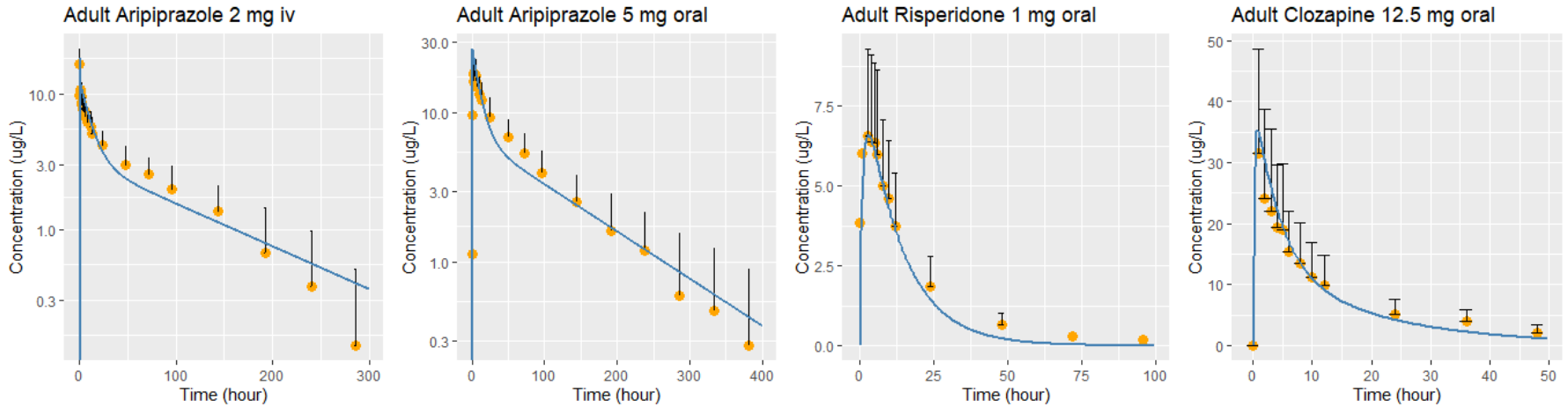
$$CL_{int, in vivo} = (CL_{int, in vitro} / FU_{mic}) \times ISEF \times MPPGL \times Abundance \times \text{liver weight}$$

In vitro to in vivo extrapolation (IVIVE)

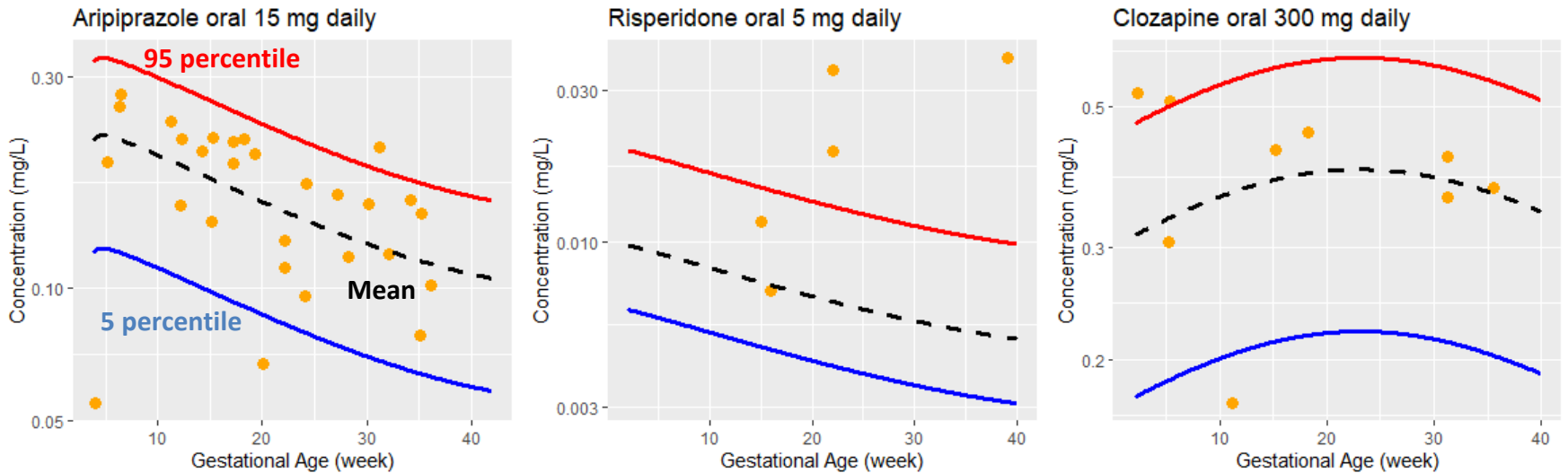
Drugs	Metabolic Enzyme Isoforms	Metabolic Pathways	Vmax	Km	CL <sub>int</sub> (in vitro)	f <sub>mic</sub> : fraction unbound in microsomal systems (if available)	ISEF: Intersystem Extrapolation Factor (if recombinant enzymes used)	MPPGL: Microsomal Protein Per Gram of Liver (life-stage specific)	Total CL <sub>int</sub> (in vivo, per of kg liver)	References
Unit	N/A	N/A	nmol/nmol P450/min	uM (pmol/μL)	μL/min/pmol	no units	unit less	mg/g	L/hr/kg	N/A
Aripiprazole	CYP2D6	DM-1451	0.42	15.7	0.027	0.435	1.069	40	26.68	FDA Review (2001)
		DM-1452	3.92	27.9	0.141	0.435	1.069	40		
		OPC-14857	2.81	26.2	0.107	0.435	1.069	40		
	CYP3A4	DM-1451	12.5	514	0.024	0.435	0.226	40		
		DM-1452	19.6	757	0.026	0.435	0.226	40		
OPC-14857	6.21	298	0.021	0.435	0.226	40				
Risperidone	CYP2D6	NA	NA	NA	7.550	1	1	40	211.92	Vieira et al. 2014
	CYP3A4	NA	NA	NA	0.180	1	1	40		
	CYP1A2	NA	13.1	14.2	0.923	1	0.571	40		
Clozapine	CYP2D6	NA	4.5	19.5	0.231	1	1.069	40	81.12	Ghoneim et al. 2020
	CYP3A4	NA	11.6	91.6	0.127	1	0.226	40		

# Model Simulation

## Model Calibration with Non-pregnant Adults



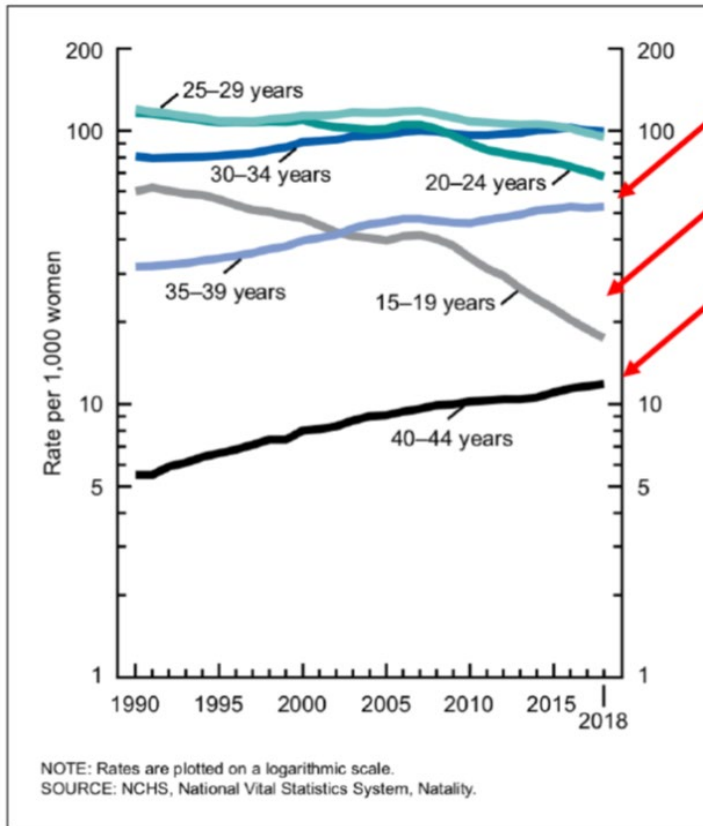
## Model Evaluation with Daily Dose through Pregnancy



# Pregnant Women Population

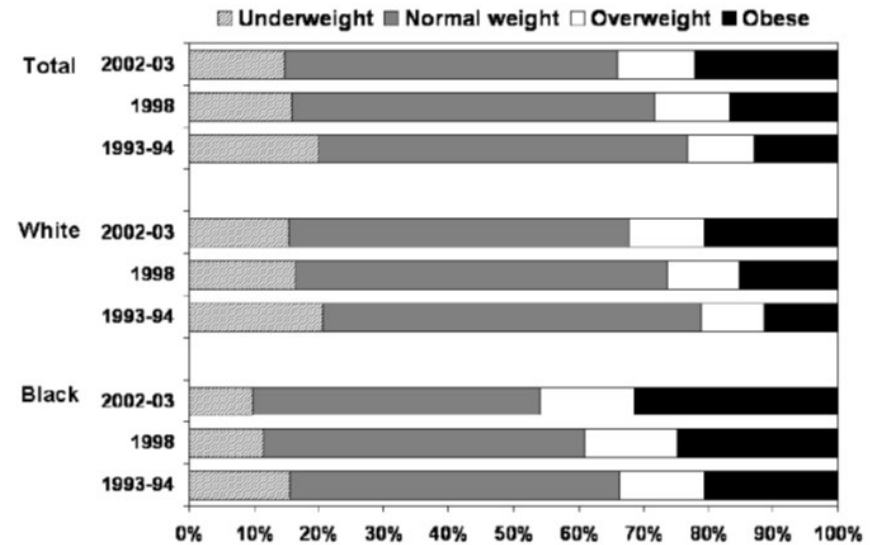
Based on National Health and Nutrition Examination Survey (NHANES)

## Age



Age ratios of pregnant women change throughout years

## Ethnicity & BMI



\*1990 cut points. SOURCE: Kim et al., 2007. (IOM Report, 2009)

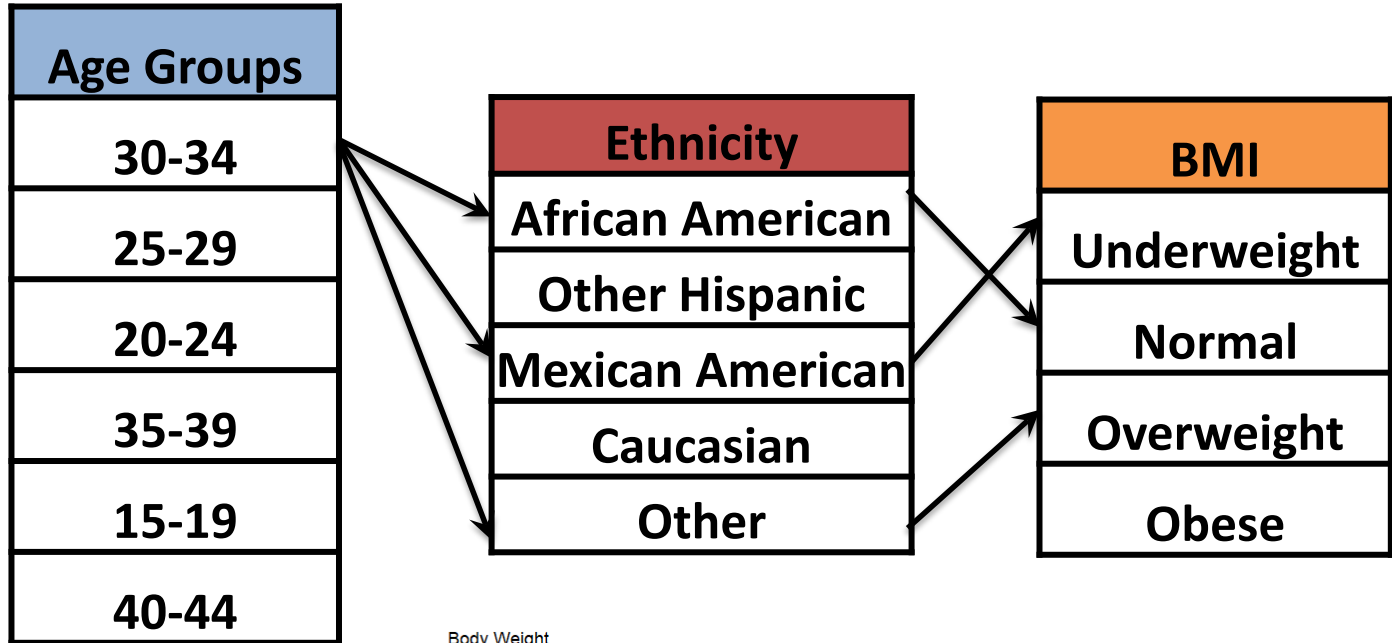
Figure 3. Birth rates, by age of mother: United States, 1990–2018

IOM Report, 2009

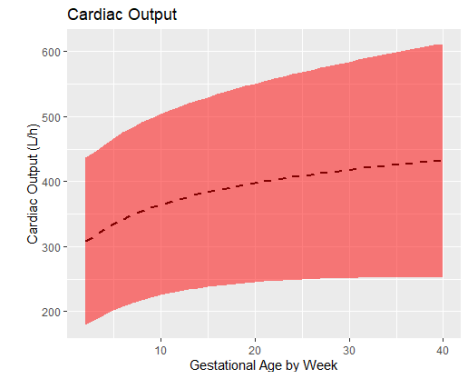
# Population Pregnancy Model

- Hierarchical Workflow (2 Steps)

1. HTTK  
Population Generation

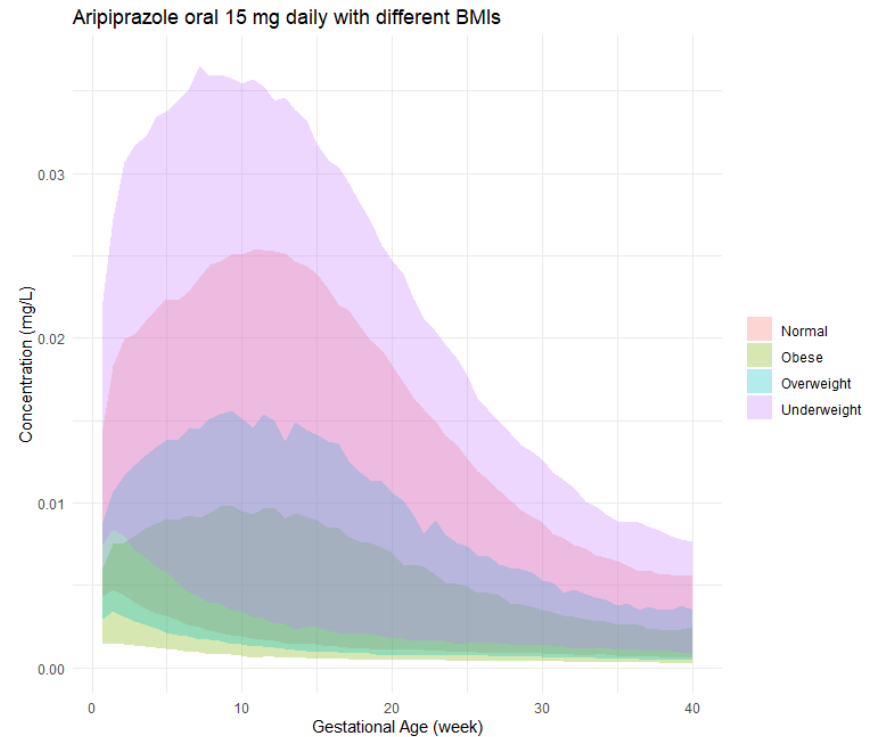
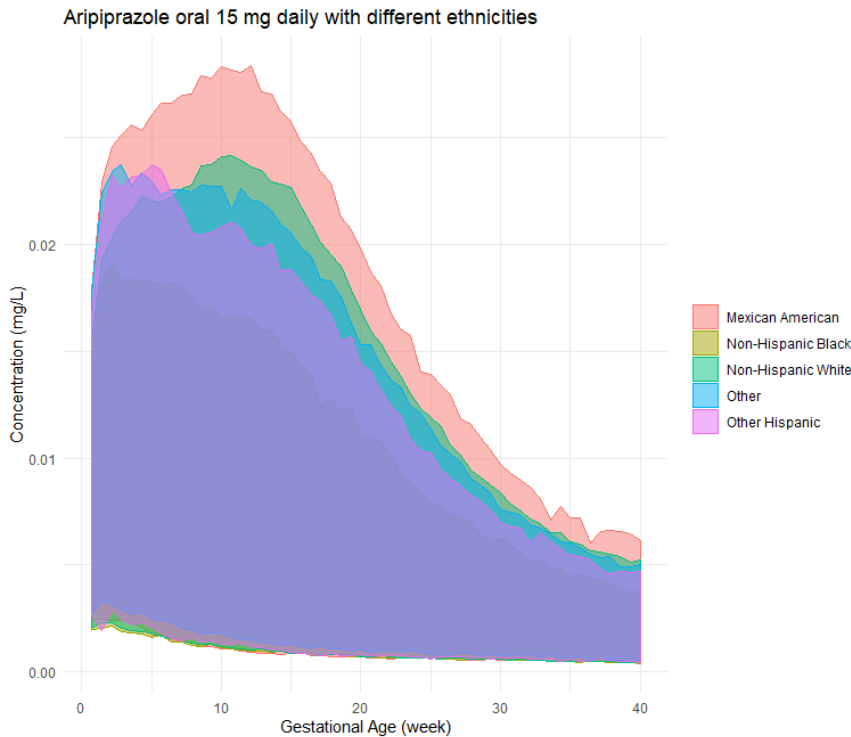


2. Monte Carlo Simulation  
Time dependent changes  
Dynamic Variability



# Pharmacokinetic Profiling

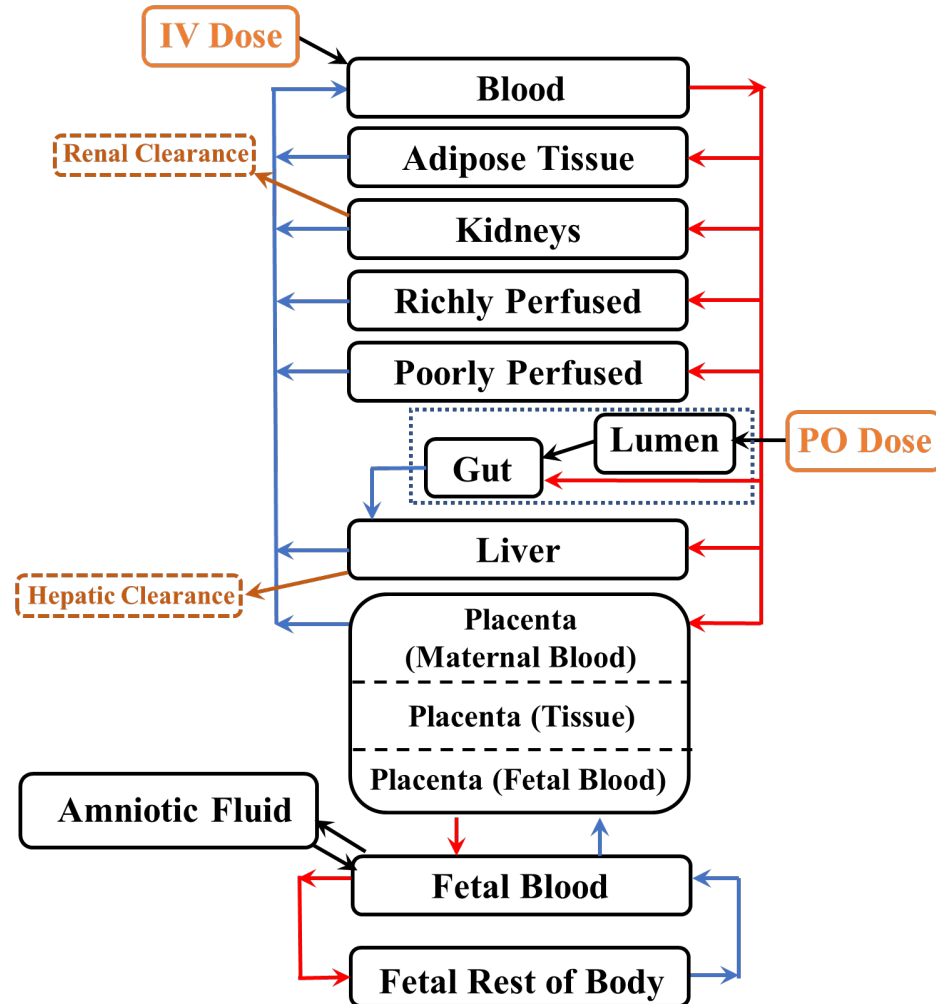
- By applying the population model based on different ethnic, BMI and age groups, the subgroups can be separated to view the PK profile individually



# Maternal-Fetal Model Simulation

## Model Assumptions:

1. The placenta considered as a permeability-limited compartment.
2. The passive diffusion of drugs is proportional to the placenta villous surface area.
3. The drug exchange between amniotic sac and maternal circulation considered negligible.
4. The drug concentrations in fetal venous plasma equals to drug concentrations in umbilical cord plasma.



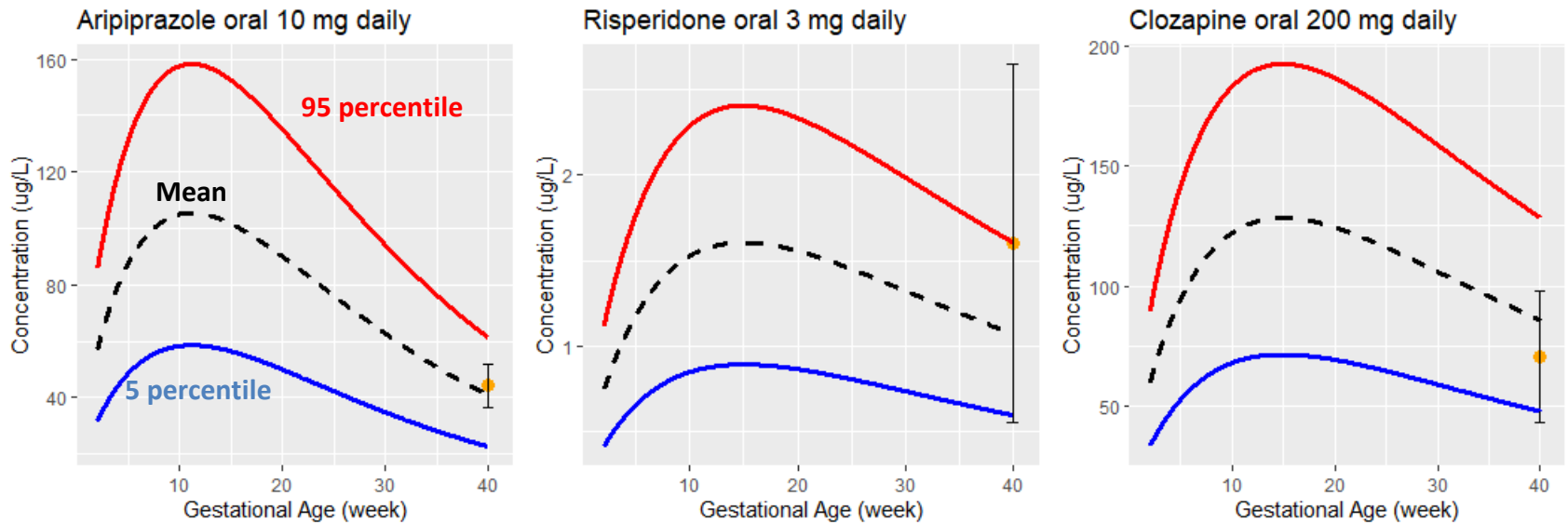


# Maternal-Fetal Model Simulation

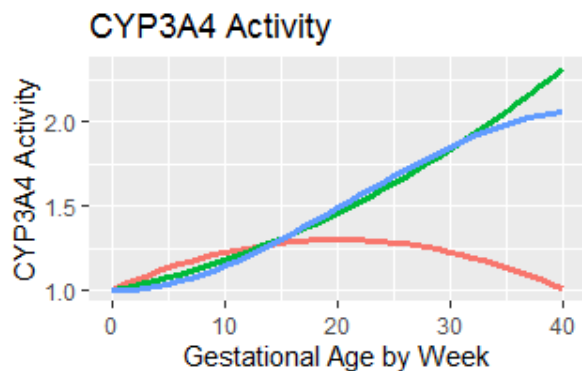
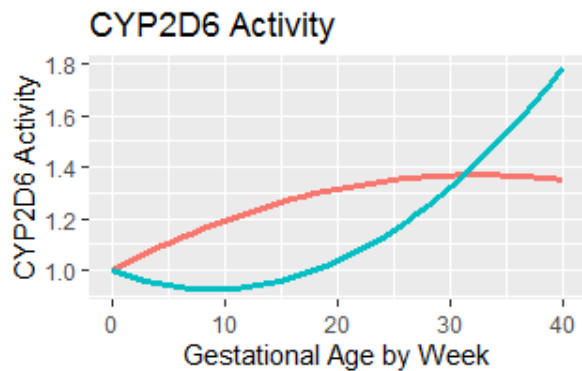
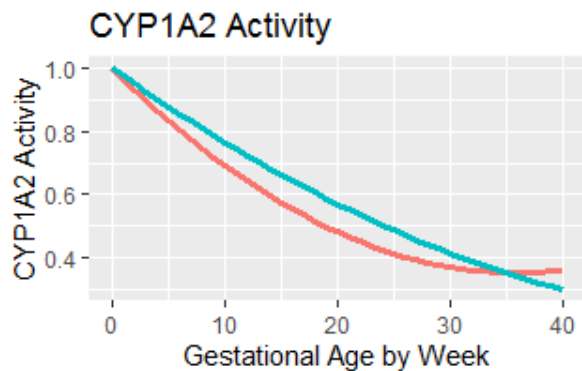
## Observed data from previous reports

	Drug Concentrations in Umbilical Cord Plasma			Dose (mg/day)	References
	n	Mean (ng/mL)	SD		
Aripiprazole	3	44.33	7.76	10 mg/day	Windhager et al. 2014
Risperidone	6	1.6	1.05	3 mg/day	Newport et al. 2007
Clozapine	5	70.4	27.4	200 mg/day	Imaz et al. 2018

## Umbilical Cord Plasma Concentrations of Antipsychotics

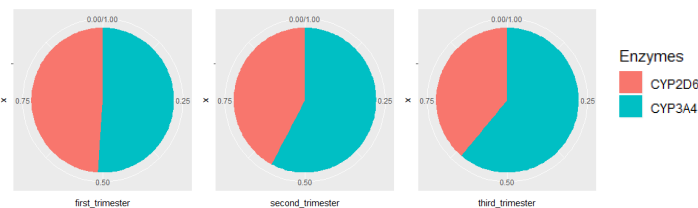
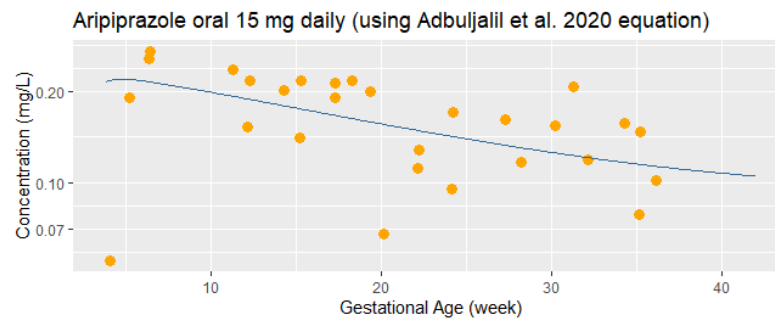
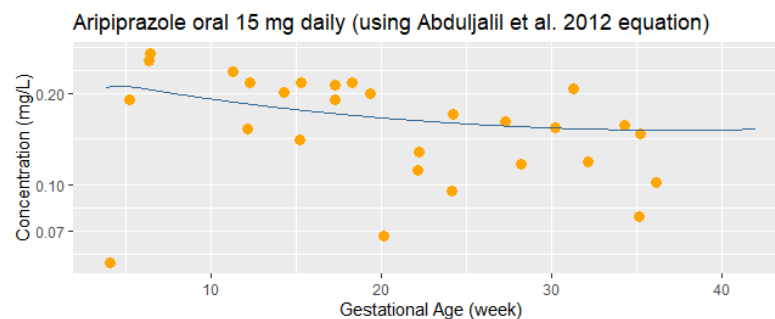
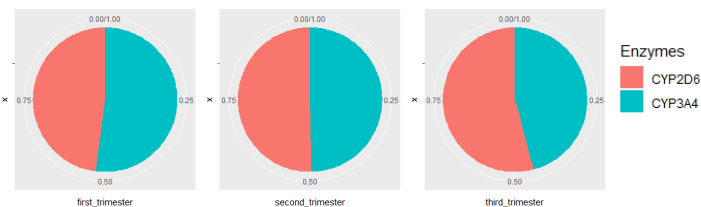


# Flexible to Keep up with Current Research



Type

- Abduljalil 2012
- Abduljalil 2020
- Ke 2019



# Conclusions

- Pregnancy and maternal-fetal PBPK is a useful tool to predict the intra-uterine exposure
- The population model based on the demographics of real population can help advance the variability assessment in these susceptible populations
- Maternal-fetal PBPK model can help determine the safety and efficacy of therapeutic agents and can also be adapted for chemical risk assessment as necessary

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