



**Drug Development Advisors**

*Driving Scientific Innovation Since 2002*

## Creating and Performing Research with PhysioPD™ Research Platforms: Process and Case Study

QSP Congress Europe 2015  
Basel, Switzerland  
28 April 2015

[www.rosaandco.com](http://www.rosaandco.com)

## Summary

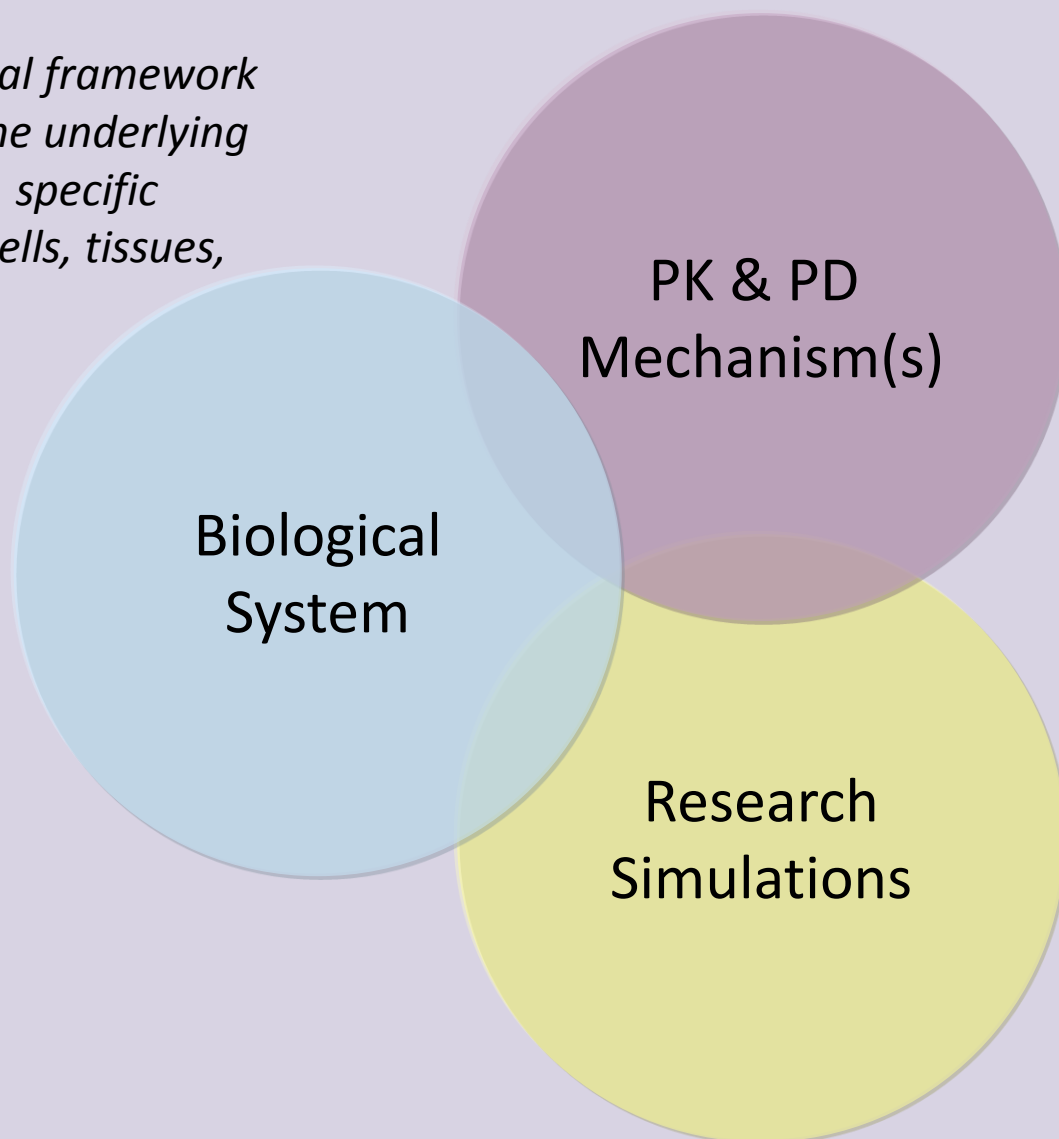
- The PhysioPD research approach is designed to impact client decisions and has been successful in many, diverse therapeutic indications
- PhysioPD Research Platforms are Quantitative Systems Pharmacology (QSP) models that are designed with multidisciplinary client team input
- I will describe the process of creating and conducting research using PhysioPD Research Platforms to drive scientific innovation in the pharmaceutical industry

PhysioPD™ Research Platforms incorporate biological mechanisms, pharmacology, and simulation capabilities.



*Mathematical framework describing the underlying biology, e.g., specific mediators, cells, tissues, organs*

*Target MOA and/or compound pharmacology*

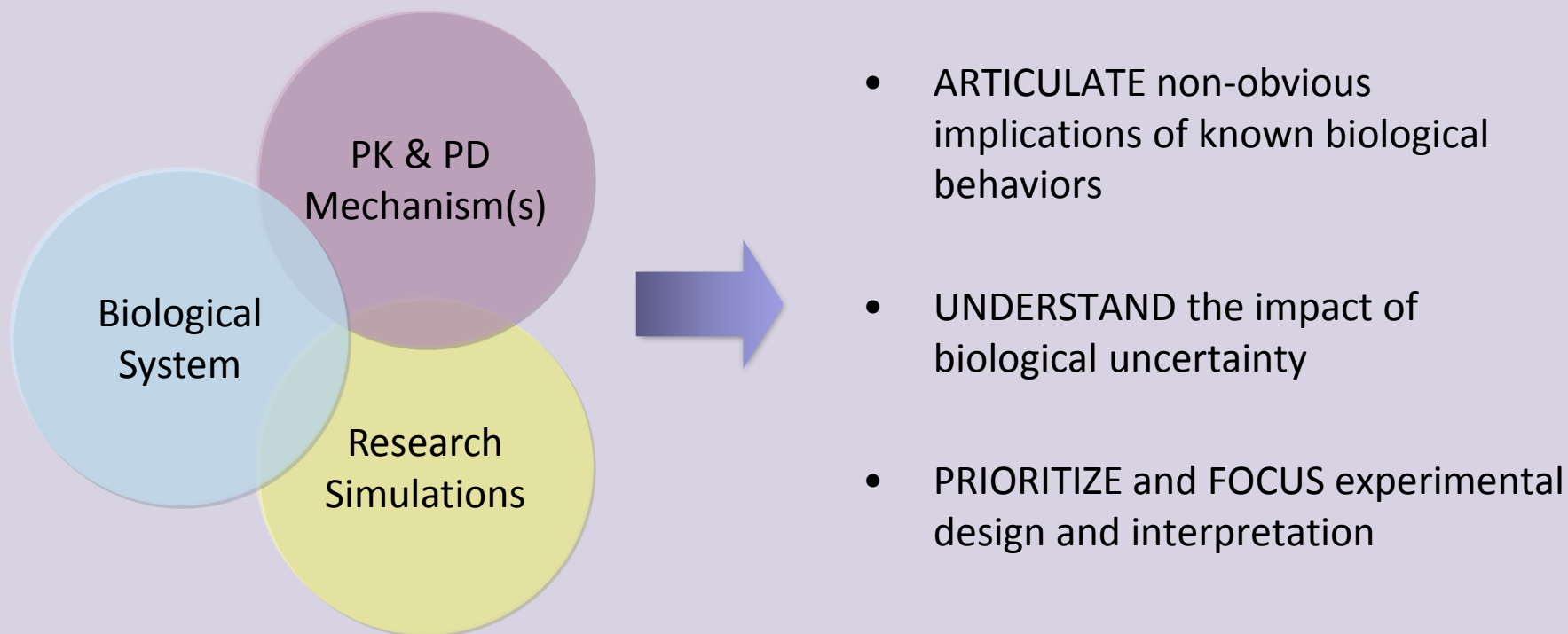


**Biological System**

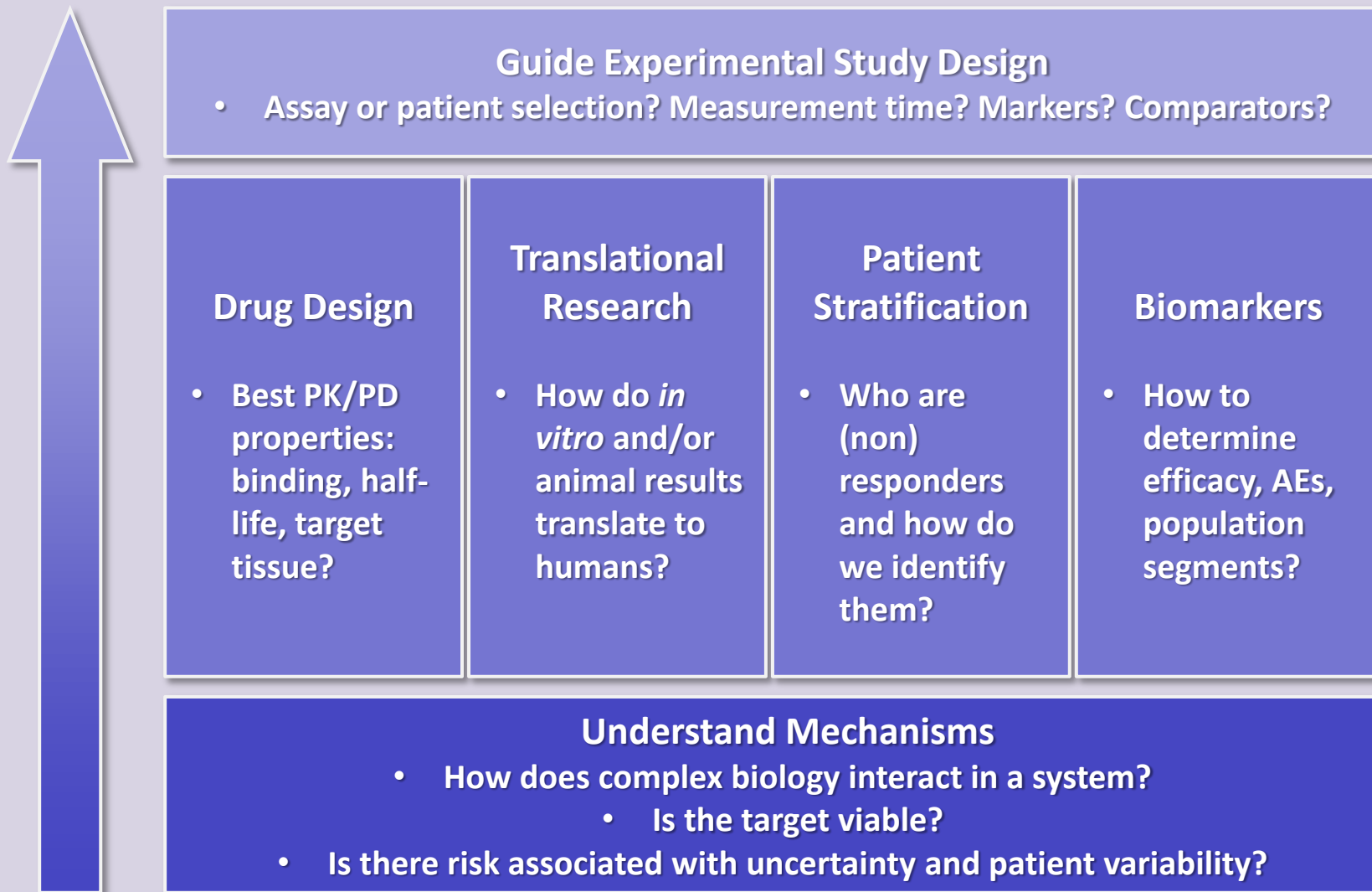
**PK & PD Mechanism(s)**

**Research Simulations**

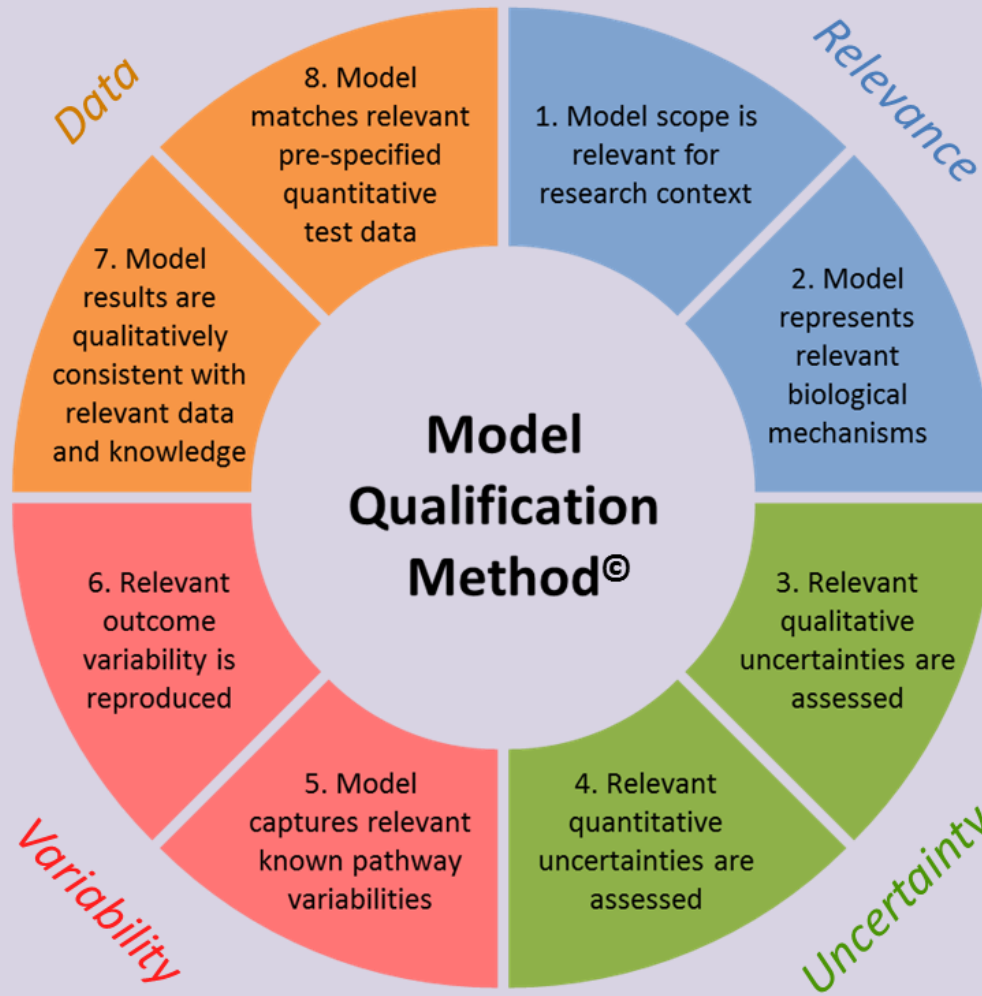
*Simulate in vitro or in vivo studies or clinical trials*



# PhysioPD Research Objectives: Connecting Mechanisms to Outcomes

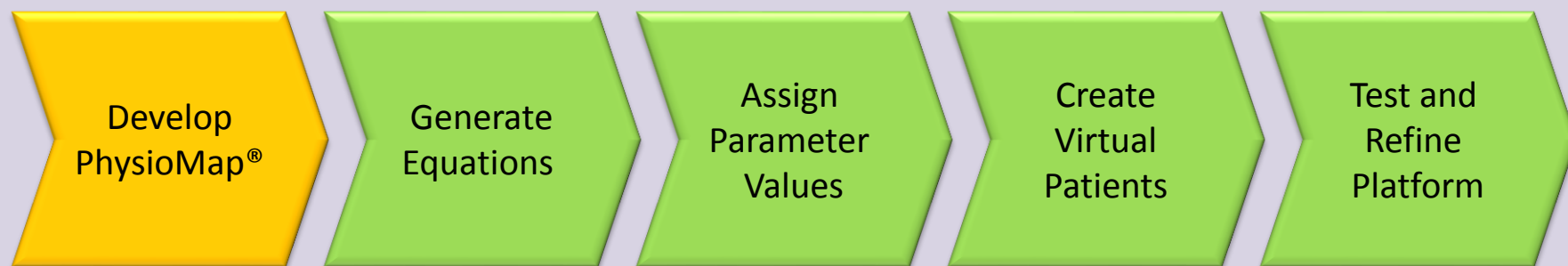


Rosa's Model Qualification Method ensures that the Platforms are fit for purpose.



Ref: Friedrich, et al. (2011)

# Process for Creating PhysioPD Research Platforms



# A PhysioPD Research Platform includes a PhysioMap® and a mathematical representation of biology: Metabolism Example



**Reaction Properties**

### Set Rate Law and Initialise Values

**Stoichiometry**

Name of Reaction:

☐ Reversible (for elementary mode calc)

Stoichiometries

1  ----> 1  + 1

☐ Use Full Names

**Kinetics**

Enter your Kinetic Rate Laws here:

Built-in Rate Laws Free Format

Type Your Rate Law Here (Hit TAB when done):

$$\text{IRR\_scale\_k} * \text{Conc\_Glucose} * \text{Insulin\_Panc\_RR} * (\text{pow}(\text{Conc\_Glucose}, \text{Glucose\_IRR\_nh}) * (1 - \text{cAMP\_Slope\_Eff})) / (\text{pow}(\text{Glucose\_IRR\_Km} - \text{cAMP\_Shift\_Eff}, \text{Glucose\_IRR\_nh}) * (1 - \text{cAMP\_Slope\_Eff})) + \text{pow}(\text{Conc\_Glucose}, \text{Glucose\_IRR\_nh}) * (1 - \text{cAMP\_Slope\_Eff})) / (1 + \text{cAMP\_Trans\_Eff}) * (1 + \text{IRR\_Ca\_k} * \text{Ca\_Eff})$$

Symbol Name	Value
IRR_scale_k	0.0003
Glucose_IRR_nh	5
Glucose_IRR_Km	120
IRR_Ca_k	0.1

Commit Parameter Changes

**Mechanism of Impaired Insulin-stimulated Muscle Glucose Metabolism in Subjects with Insulin-dependent Diabetes Mellitus**

Gary W. Cline, Inger Magnusson, Douglas L. Rothman, Kitt Falk Petersen, Didier Laurent, and Gerald I. Shulman  
Department of Internal Medicine, Yale University School of Medicine, New Haven, Connecticut 06520

**Abstract**

To determine the mechanism of impaired insulin-stimulated muscle glucose metabolism in subjects with insulin-dependent diabetes mellitus (IDDM), we used <sup>13</sup>C-NMR spectroscopy to measure the C1 resonance of the glucose during a 6-h hyperglycemic-hyperinsulinemic (G-6-P) clamp in 10 IDDM subjects and 10 age-matched control subjects. The rate of glucose uptake (μM/min/100 ml forearm) was significantly lower in the IDDM subjects (1.95 ± 0.6) than in the control subjects (3.0 ± 0.5, P < 0.05). The reduction in the rate of glucose uptake was not due to differences in the rate of glucose infusion (1.95 ± 0.6 vs. 3.0 ± 0.5 μM/min/100 ml forearm). When nonoxidative glucose disposal was estimated by the rate of glucose disappearance (Rd), the rate of Rd was also lower in the IDDM subjects (0.07 ± 0.02 vs. 0.15 ± 0.02 μM/min/100 ml forearm, P < 0.05). These data indicate that the rate of glucose uptake is impaired in IDDM subjects. When nonoxidative glucose disposal was estimated by the rate of glucose disappearance (Rd), the rate of Rd was also lower in the IDDM subjects (0.07 ± 0.02 vs. 0.15 ± 0.02 μM/min/100 ml forearm, P < 0.05). These data indicate that the rate of glucose uptake is impaired in IDDM subjects.

**Graphical Output**

Edit Graph x: 128.83 Load External Data X Axis: Time

conc\_Glucose (red line)  
conc\_Insulin (blue line)

Time

concentration (μM/min/100 ml forearm)

JDesigner can be obtained at <http://jdesigner.sourceforge.net/Site/JDesigner.html>



PhysioPD Research Platforms are built with extensive research, curation and integration of disparate information.



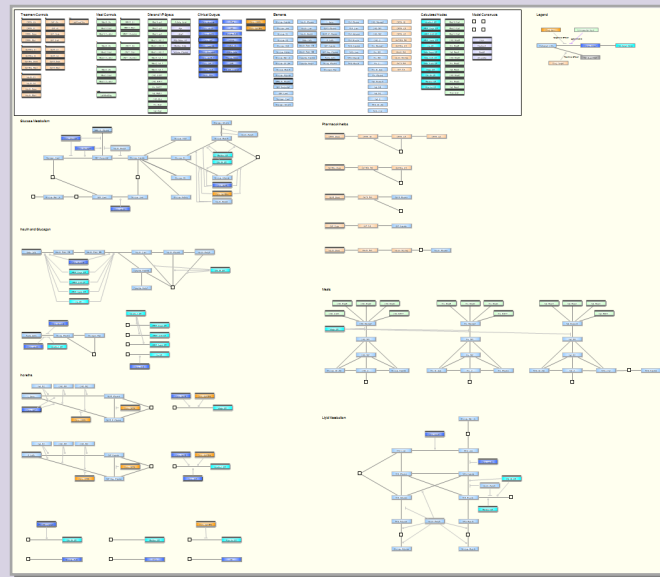
Physical Laws

Healthy & Disease  
Physiology

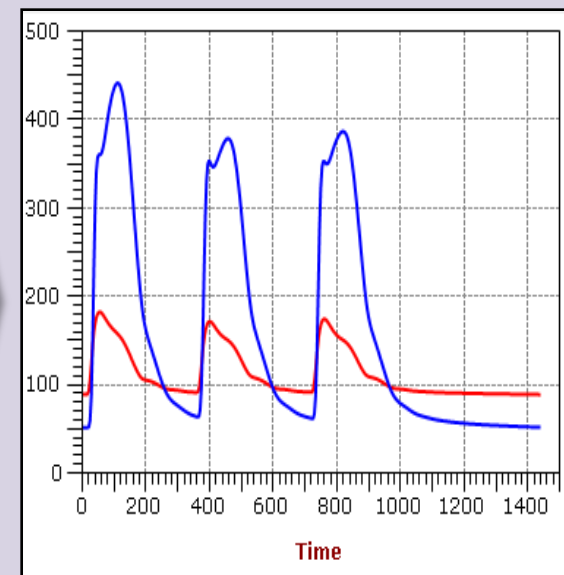
Target(s) & Drug  
Mechanism(s)

Preclinical  
Pharmacology

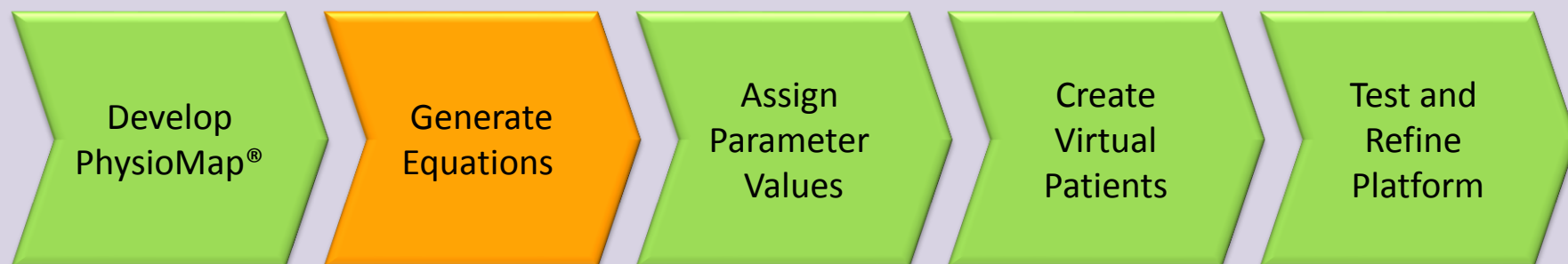
## PhysioPD Research Platform



## PhysioPD Research Results



# Process for Creating PhysioPD Research Platforms



# Rate arrows in a Platform are quantified using standard engineering techniques to represent biological interactions.

Examples of common equation forms:

- First Order Equations

$$rate\_k * S$$

- Hill Equation Modifier – Potentiation

$$1 + Emax \times \frac{L^{nh}}{EC50^{nh} + L^{nh}}$$

- Hill Equation Modifier – Activation

$$Emax \times \frac{L^{nh}}{EC50^{nh} + L^{nh}}$$

- Hill Equation Modifier - Inhibition

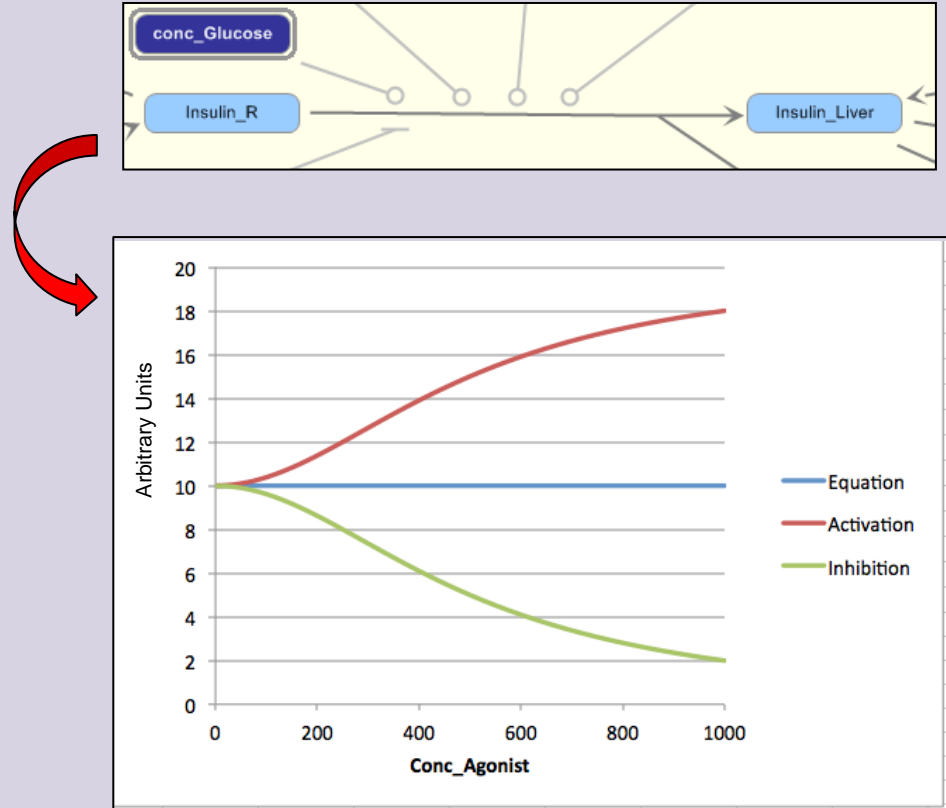
$$1 - Imax \times \frac{L^{nh}}{IC50^{nh} + L^{nh}}$$

Emax, Imax = maximum activation or inhibition effect (Emax ≥ 0, 0 ≤ Imax ≤ 1)

L = amount of ligand present

EC50, IC50 = ligand amount at 50% effect

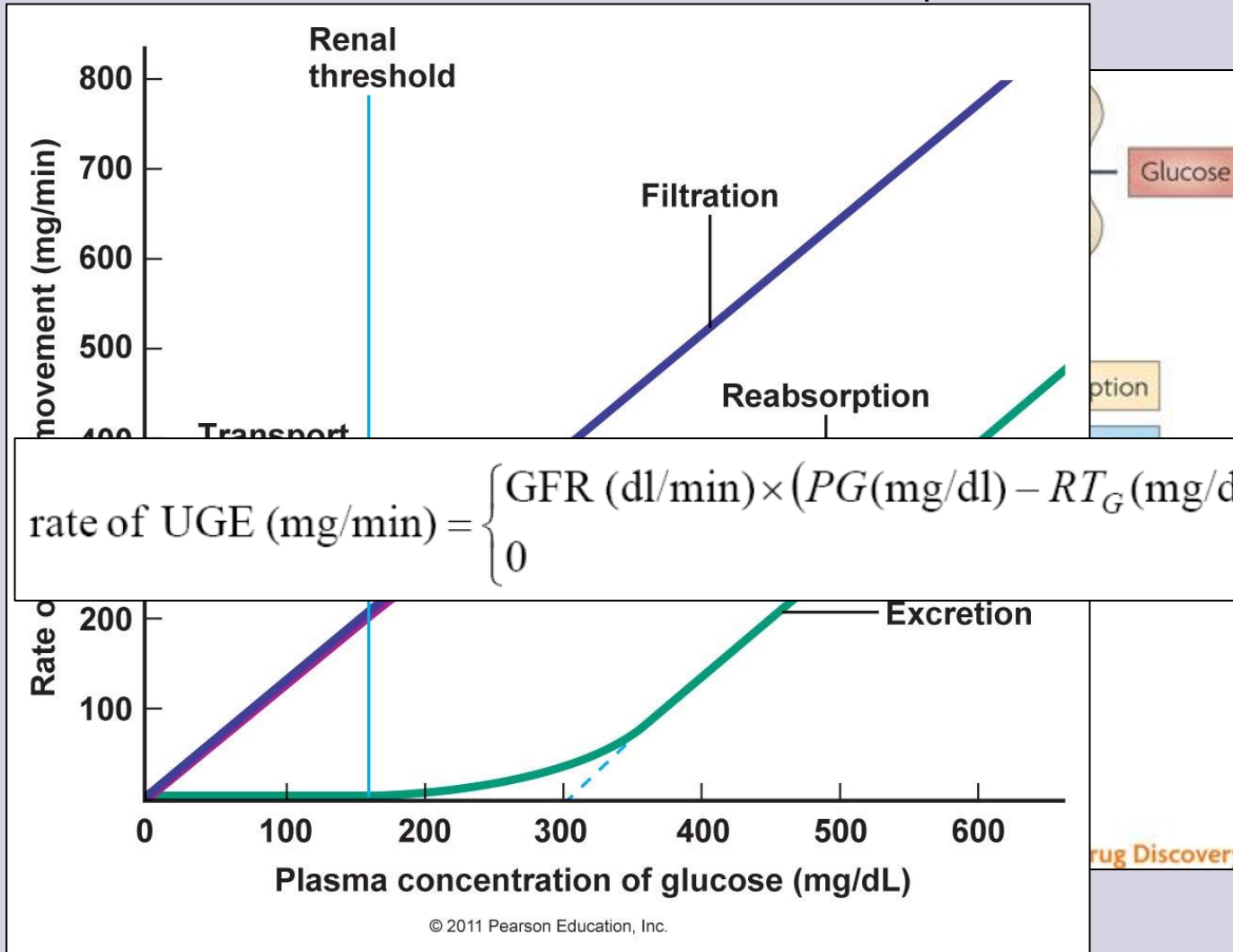
nh = Hill coefficient



Example: modeling of mediator effects.

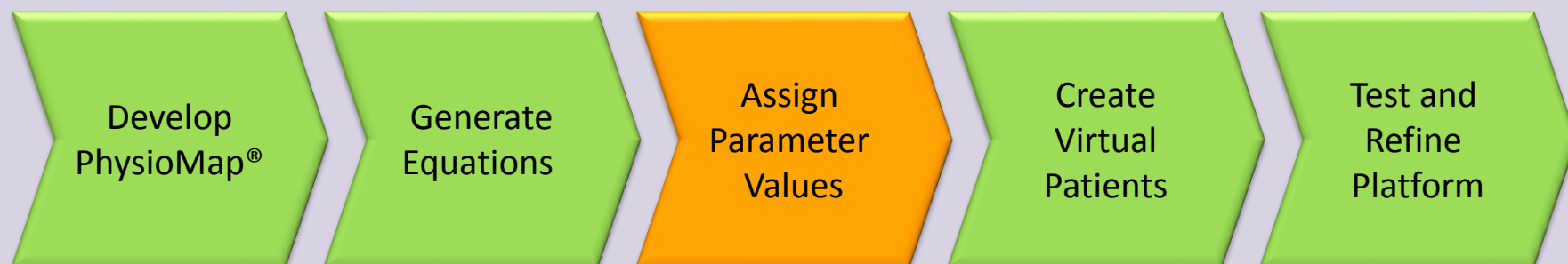
Equation forms may be derived from first principles, locally fitted to mechanistic data, or created by hypothesis.

## Renal Glucose Reabsorption



$$\text{rate of UGE (mg/min)} = \begin{cases} \text{GFR (dl/min)} \times (PG(\text{mg/dl}) - RT_G(\text{mg/dl})) & \text{if } PG > RT_G \\ 0 & \text{if } PG \leq RT_G \end{cases}$$

# Process for Creating PhysioPD Research Platforms

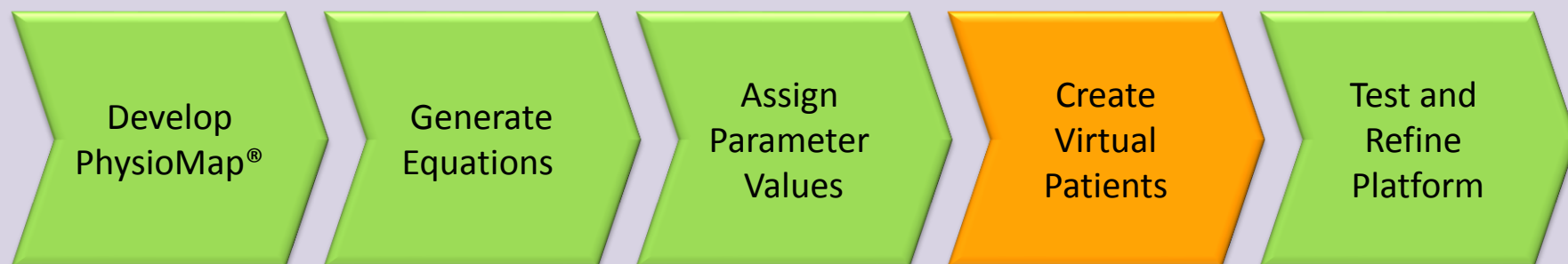


Parameter values in a Platform are identified by literature survey and data analysis, local fitting, or hypotheses.



Reference	Description	Tissue	Disease status	Type or Specie	Amount	Units	Amount in Model Units	Model Units
<b>Insulin kinetics</b>								
<b>Insulin clearance</b>								
Tura 2001	hepatic insulin extraction	liver	healthy	human	41.3	L/min	41.3	L/min
Tura 2001	Hepatic insulin clearance	liver	healthy	human	0.66	L/min	0.66	L/min
Sherwin et al., 1974	<b>Description</b>	<b>Tissue</b>	<b>Disease status</b>	<b>Type or Specie</b>	<b>Amount</b>	<b>Units</b>		
Sherwin et al., 1974								
Polonsky 1988								
Tura 2001								
Sherwin et al., 1974								
Sherwin et al., 1974								
Sherwin et al., 1974								
Krützfeldt 2000								
Sherwin et al., 1974								
	in extraction	liver	healthy	human	41.3	L/min		
	in clearance	liver	healthy	human	0.66	L/min		
<b>Insulin</b>	tion hepatic	liver	healthy	human	47	%		
<b>Not healthy</b>	nce hepatic	liver	healthy	human	400	ml/min		
Tura 2001	c insulin extraction	liver	healthy	human	53.1	%		
Tura 2001	ulin clearance	whole body	healthy	human	1.19	L/min		
Polonsky 1988	tion peripheral	Whole body	healthy	human	20	%		
Tura 2001	asma flow	Whole body	healthy	human	660	ml/min		

## Process for Creating PhysioPD Research Platforms



Specific parameters in a Platform are adjusted to create Virtual Patients (VPs) with different pathophysiology or phenotypes.

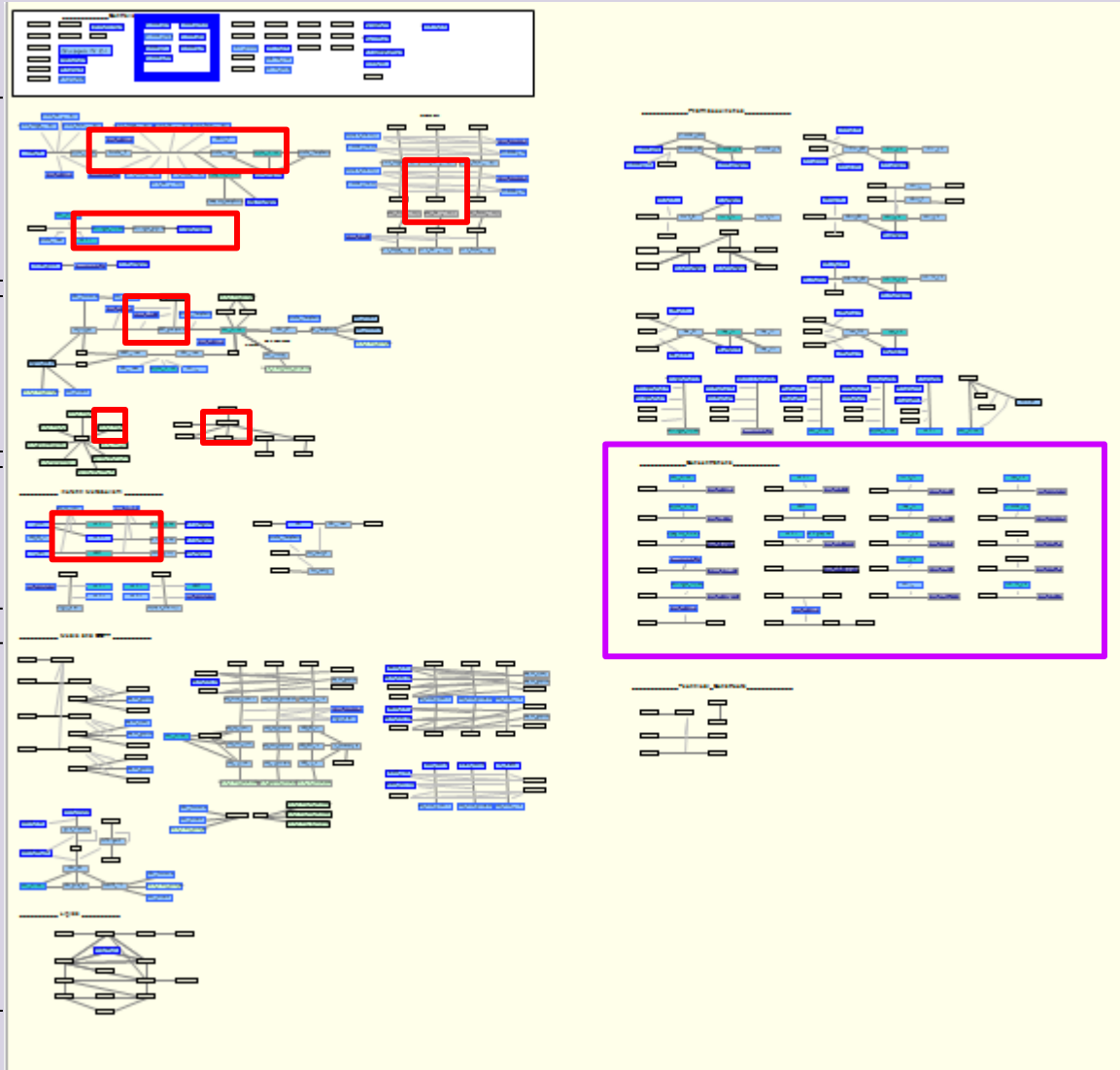


Pancreatic  
function

Glucose  
metabolism

Incretin  
production

Meal inputs  
and OGTT



Pharmacokinetics

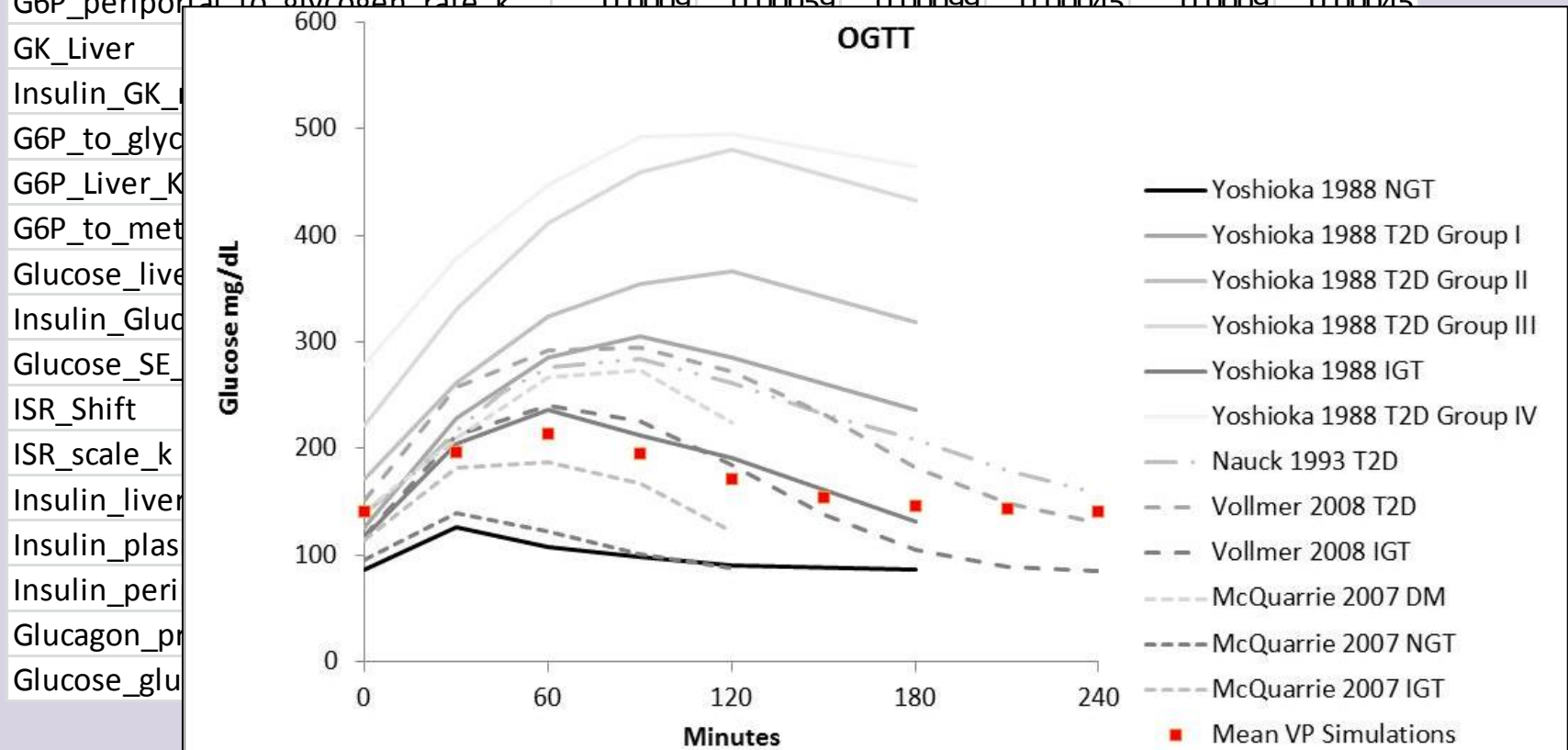
Clinical  
Outcomes



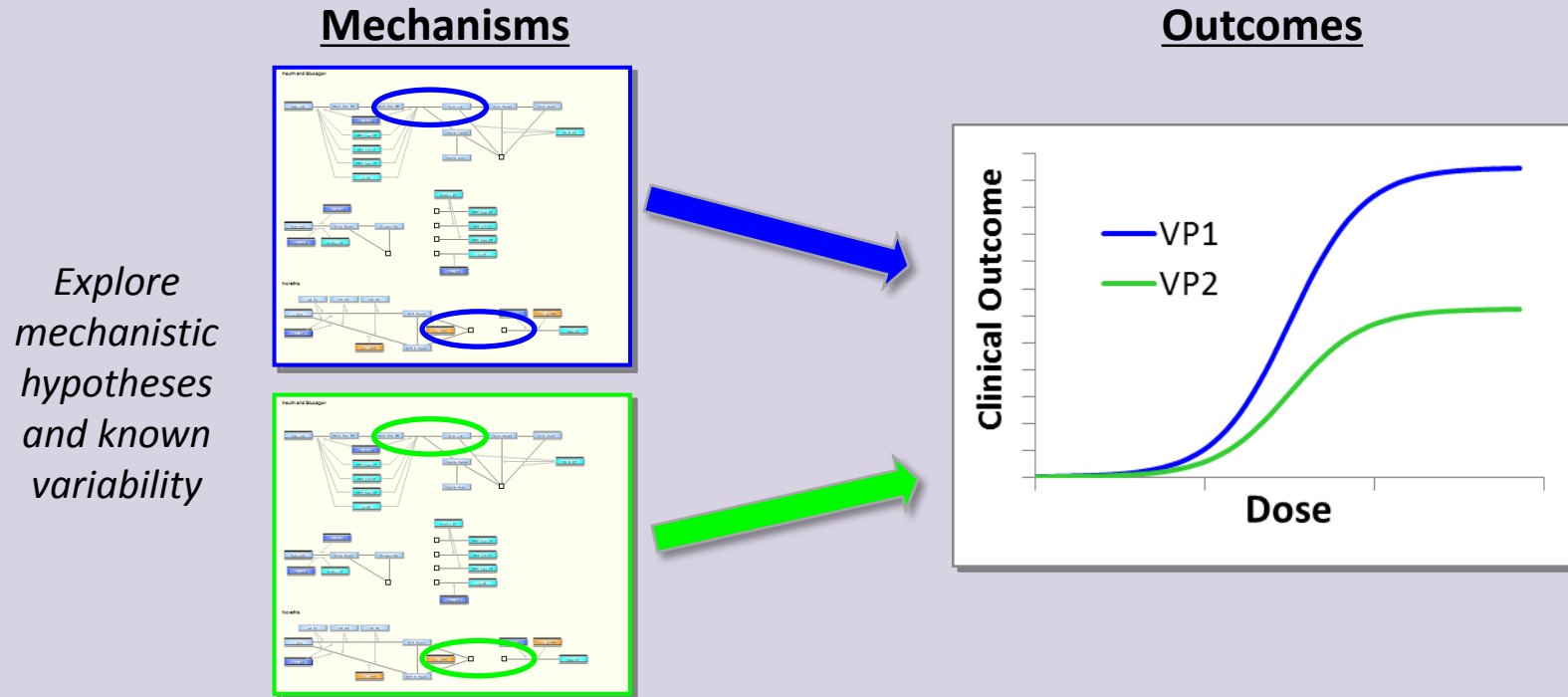
Alternate VPs are created with biologically plausible parameter values that are constrained by data and system behaviors.



	Healthy	VP1	VP2	VP3	VP4	VP5
Gluconeog_rate_max	106.9	116.9	126.9	185	126.9	250
Gluconeog_Shift	90	110	95	135	101	210
Glycogenolysis_rate_k	1.6	1.8	2.1	1.5	1.6	2.555
Body_weight	70	127	75.3	108.9	81	123
G6P_periportal_to_glycogen_rate_k	0.0009	0.00059	0.00099	0.00045	0.0009	0.00045

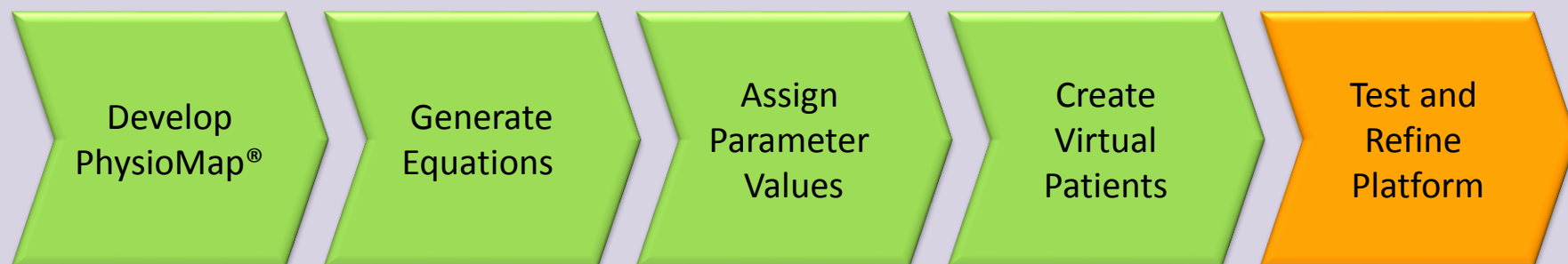


VPs facilitate exploration of how mechanistic biological differences may affect clinical outcomes.

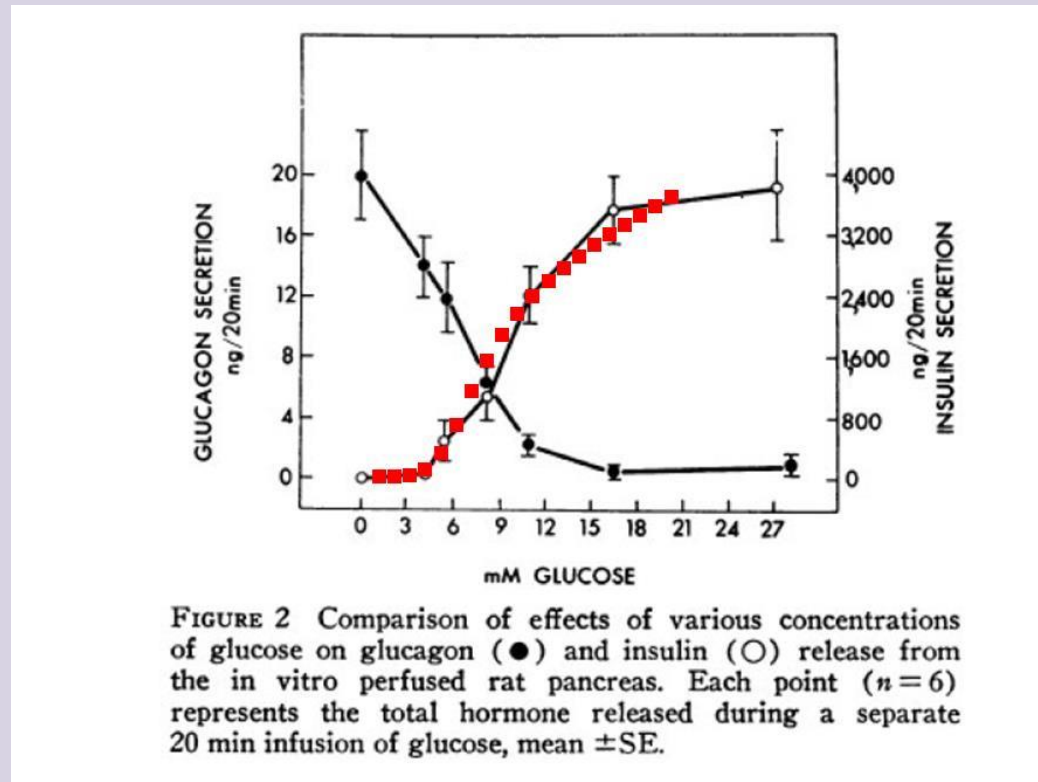


- What type of patient is most likely to respond well?
- What biomarkers are most informative?
- What enrollment criteria or protocol optimizes chances of clinical success?

# Process for Creating PhysioPD Research Platforms

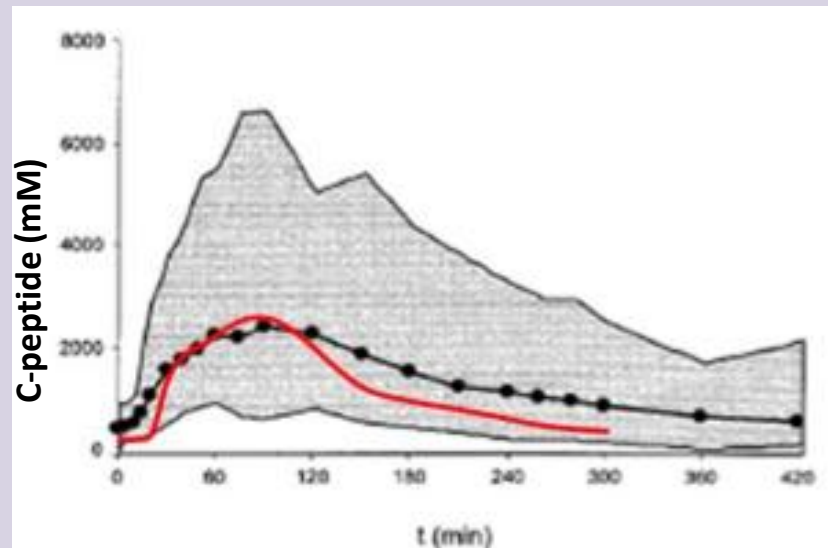
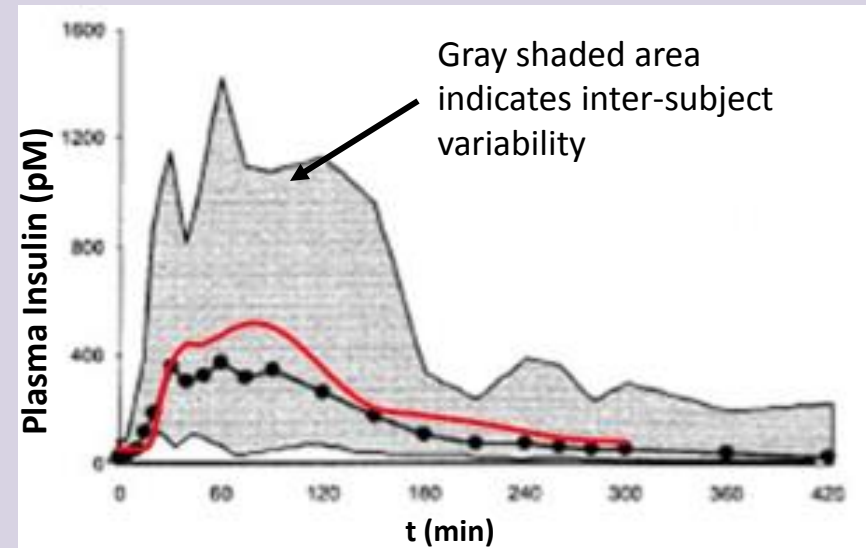
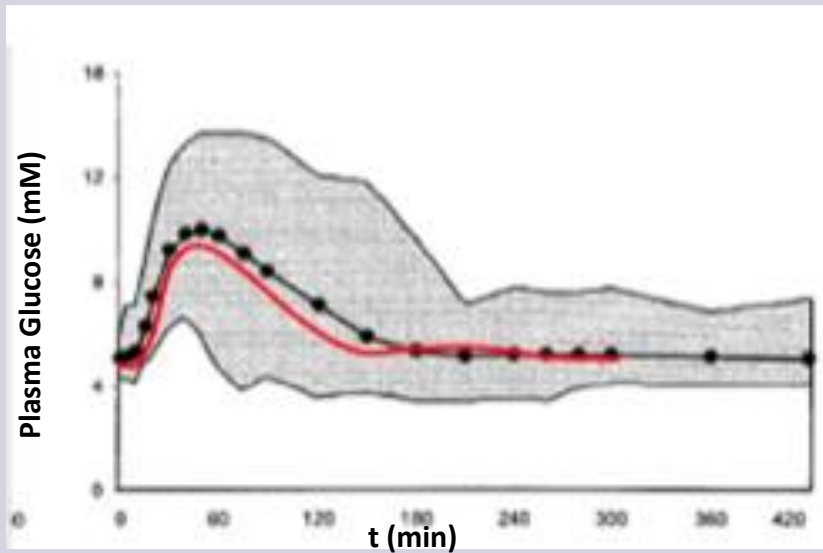


A Platform is tested against multiple datasets describing sub-system behaviors and refined if necessary.



- The simulated insulin secretion rate as a function of glucose concentration (red squares) is in agreement with experimental data (Gerich, et al. 1974)

A Platform is then tested against multiple datasets describing whole-system behaviors and refined if necessary.



- Red line is simulation
- Data from Dalla Man (2005)



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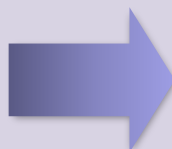
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## Case Study

## Critical questions for a program entering clinical development

- Target evaluation
  - Will a compound against this drug target be efficacious in humans?
  - Which mechanisms of action are critical for efficacy?
- Translational medicine
  - Are our preclinical data predictive of efficacy in humans?
- Clinical trial optimization
  - How will different types of patients respond to the compound?
  - Can we prospectively identify patients likely to respond?
  - What is the most efficient trial design to demonstrate treatment effects?

Mechanisms



Outcomes

A Disease PhysioMap represented the key aspects of the biology relevant to type 2 diabetes and the research questions.

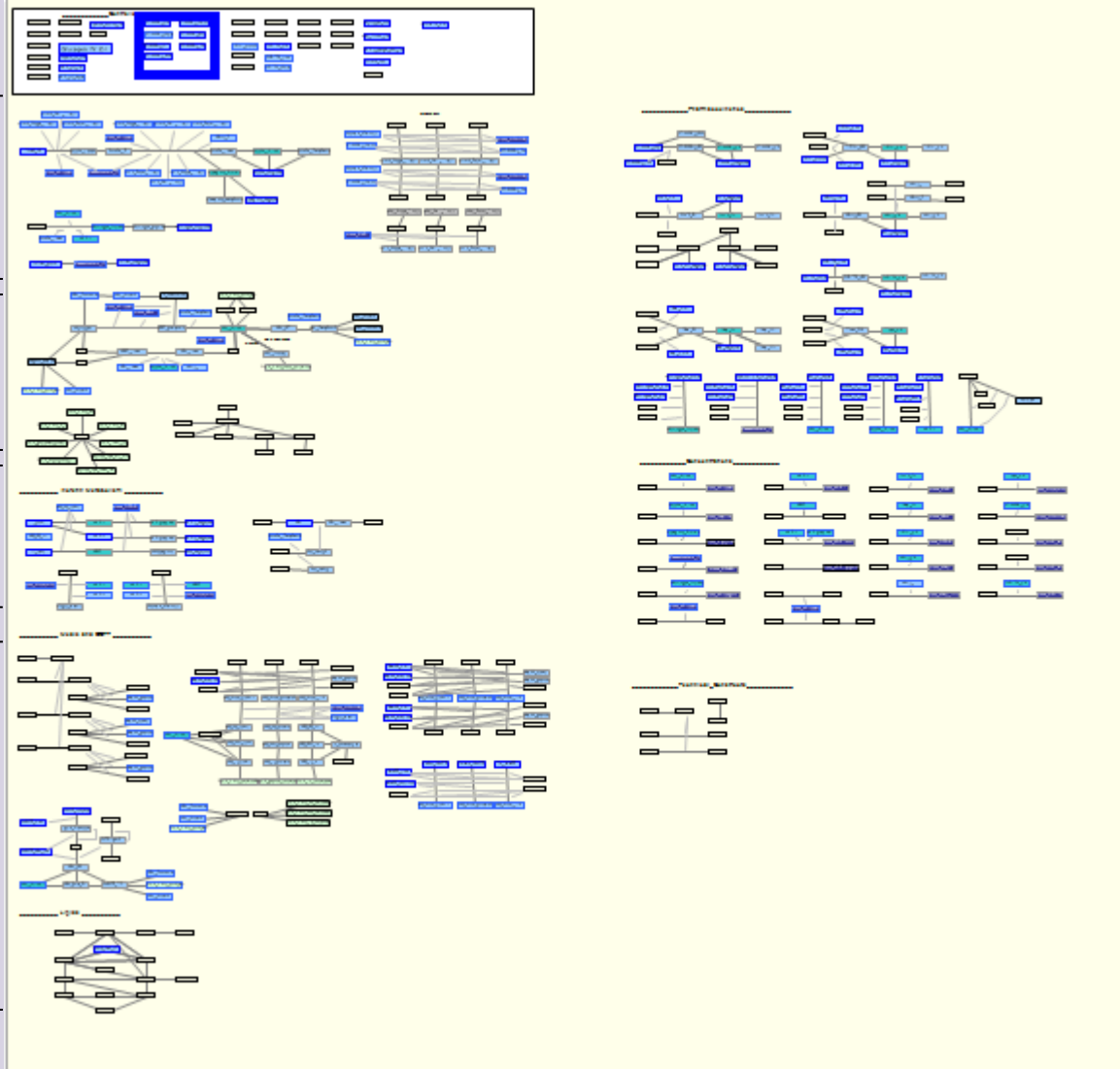


Pancreatic  
function

Glucose  
metabolism

Incretin  
production

Meal inputs  
and OGTT



Pharmacokinetics

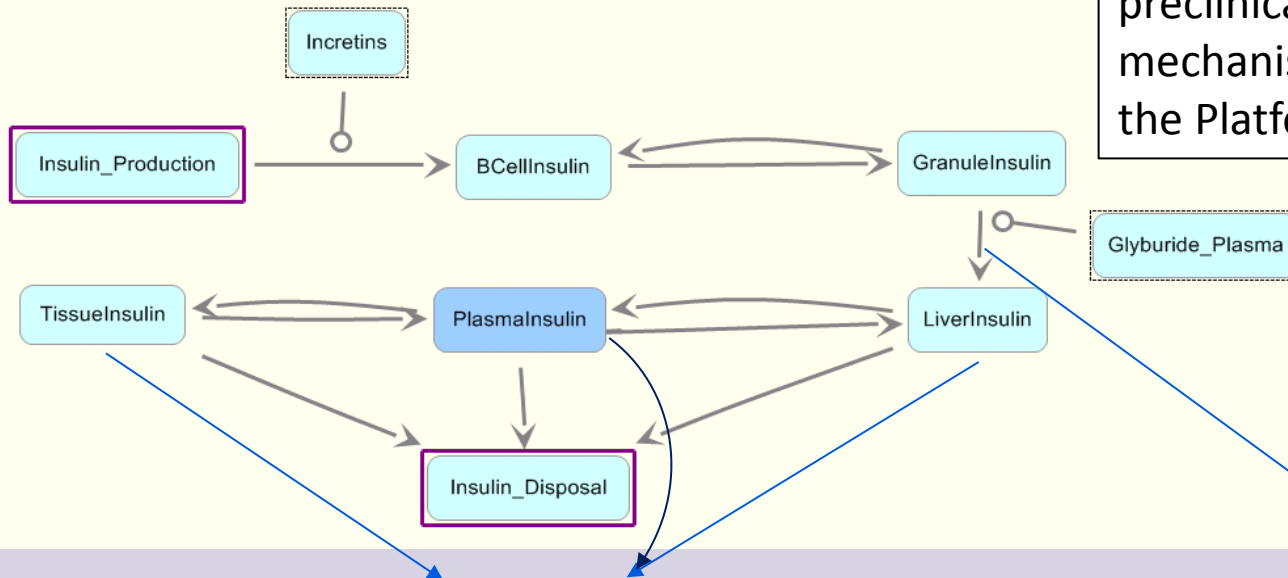
Clinical  
Outcomes



# A PhysioPD Platform represented the quantitative relationships between elements of the biological system.

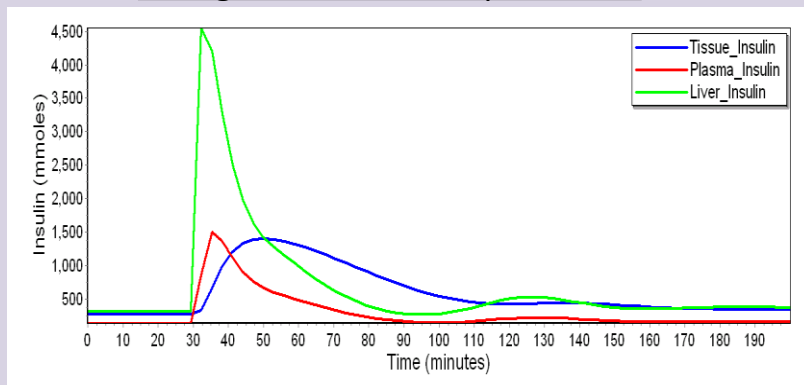


## Pancreas

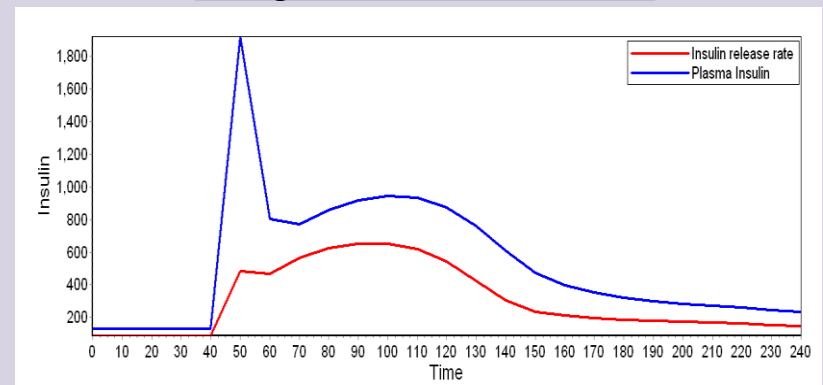


Published and proprietary preclinical data provided key mechanistic information to build the Platform.

Changes in insulin compartments

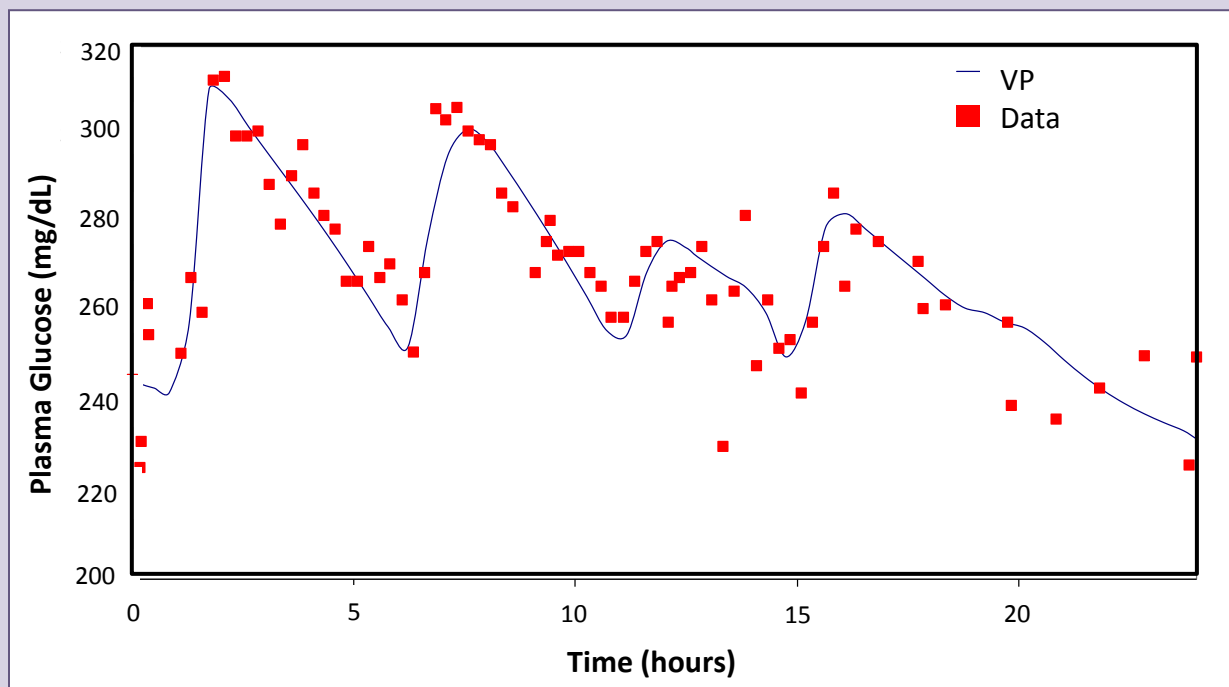


Changes in insulin release rate



VPs were created to simulate clinical trials with the compound.

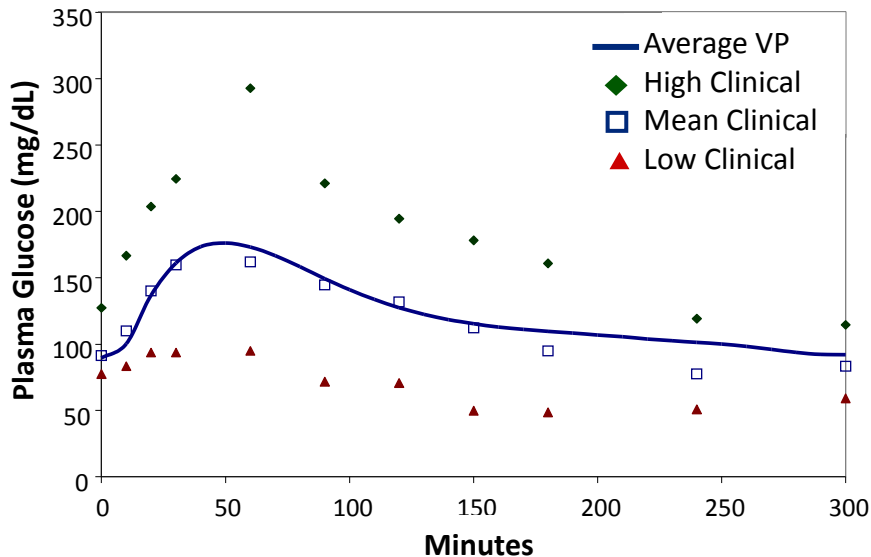
Single Virtual Patient compared to data



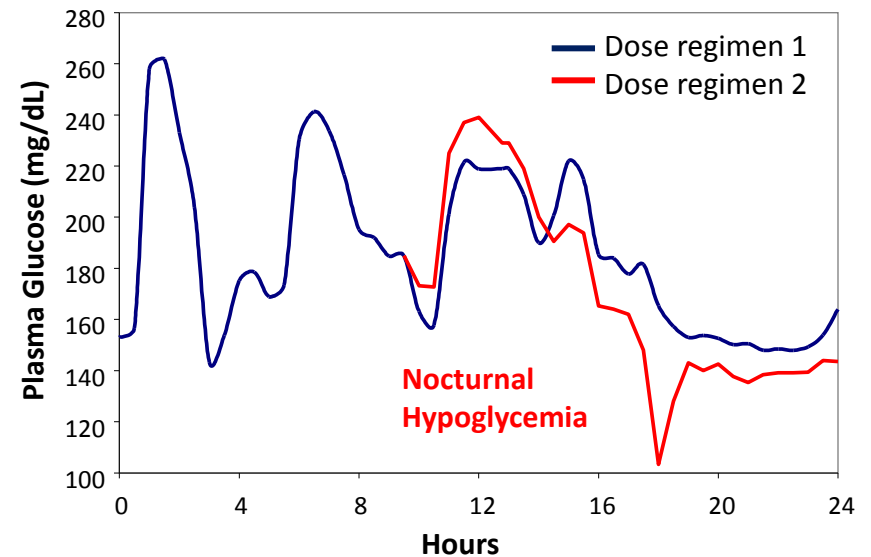
A wide range of protocols under consideration were simulated to guide the design of the clinical trial.



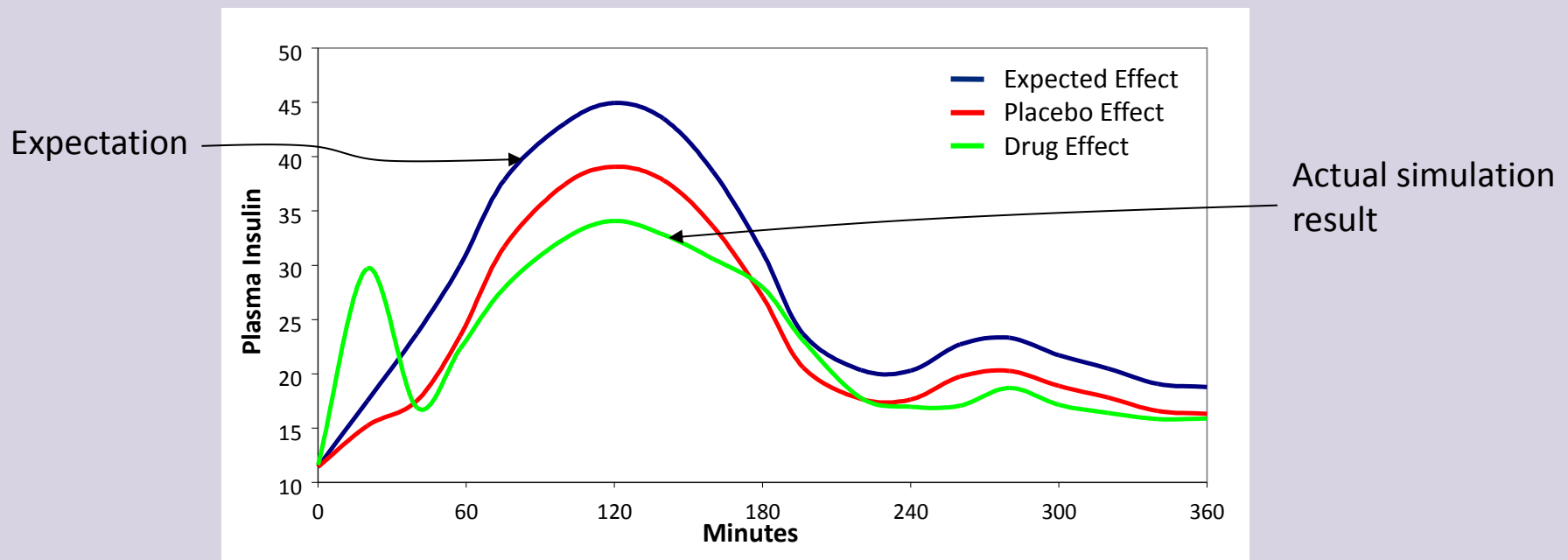
Ave. model OGTT compared to public literature



Multiple meals, snacks, complex dosing

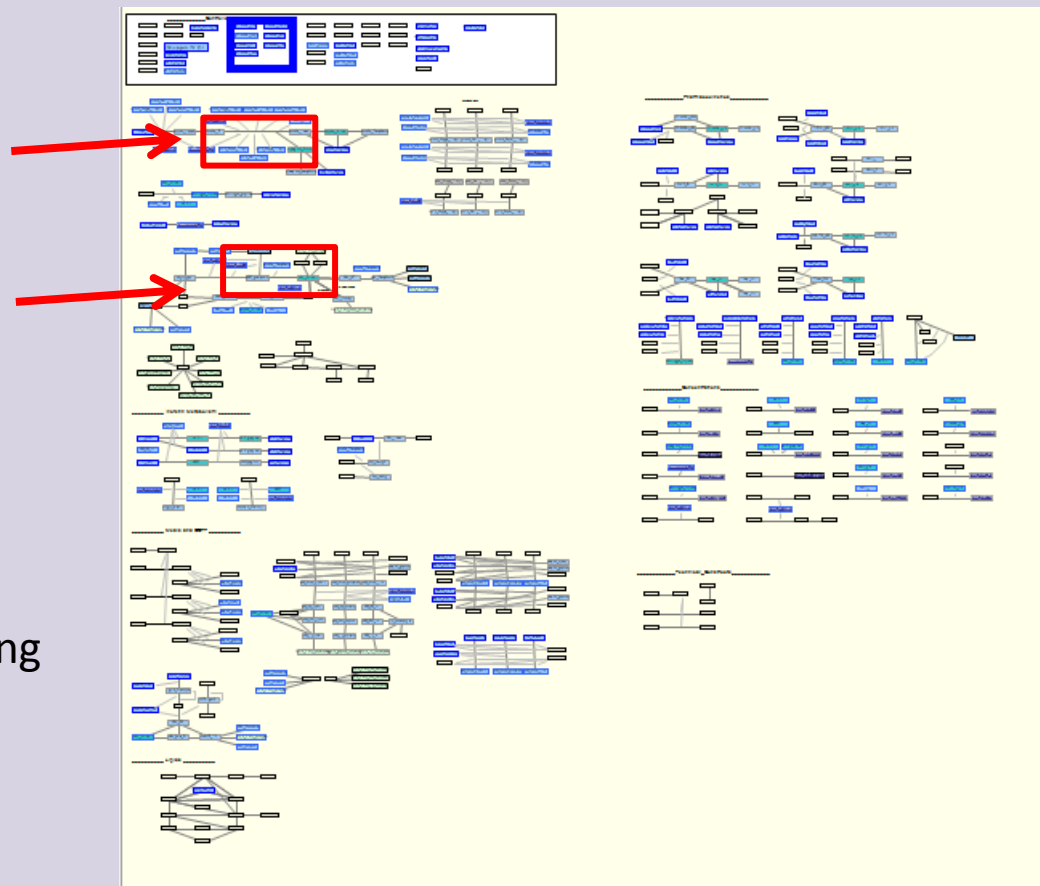
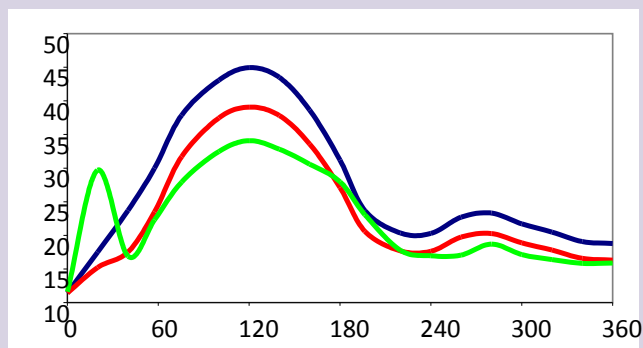


Contrary to client expectations, PhysioPD research showed that compound administration would lower plasma insulin.



PhysioPD research provided a mechanistic rationale for the unexpected behavior of the compound.

- The PhysioMap process identified multiple hypothesized compound effects
- These effects have opposite effects on insulin secretion

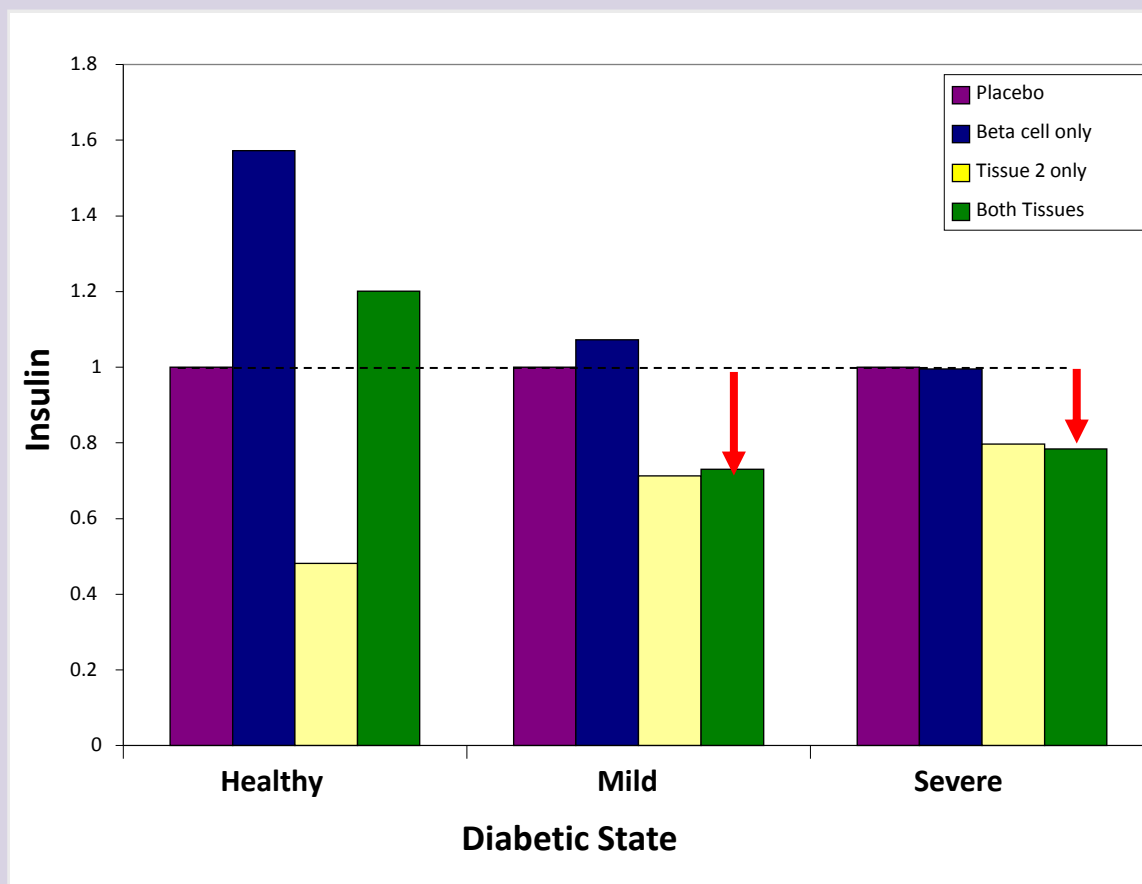


- This complex behavior was not previously identified using non-mechanistic PK/PD modeling

Simulations highlighted the relative impact of each hypothesized compound effect.



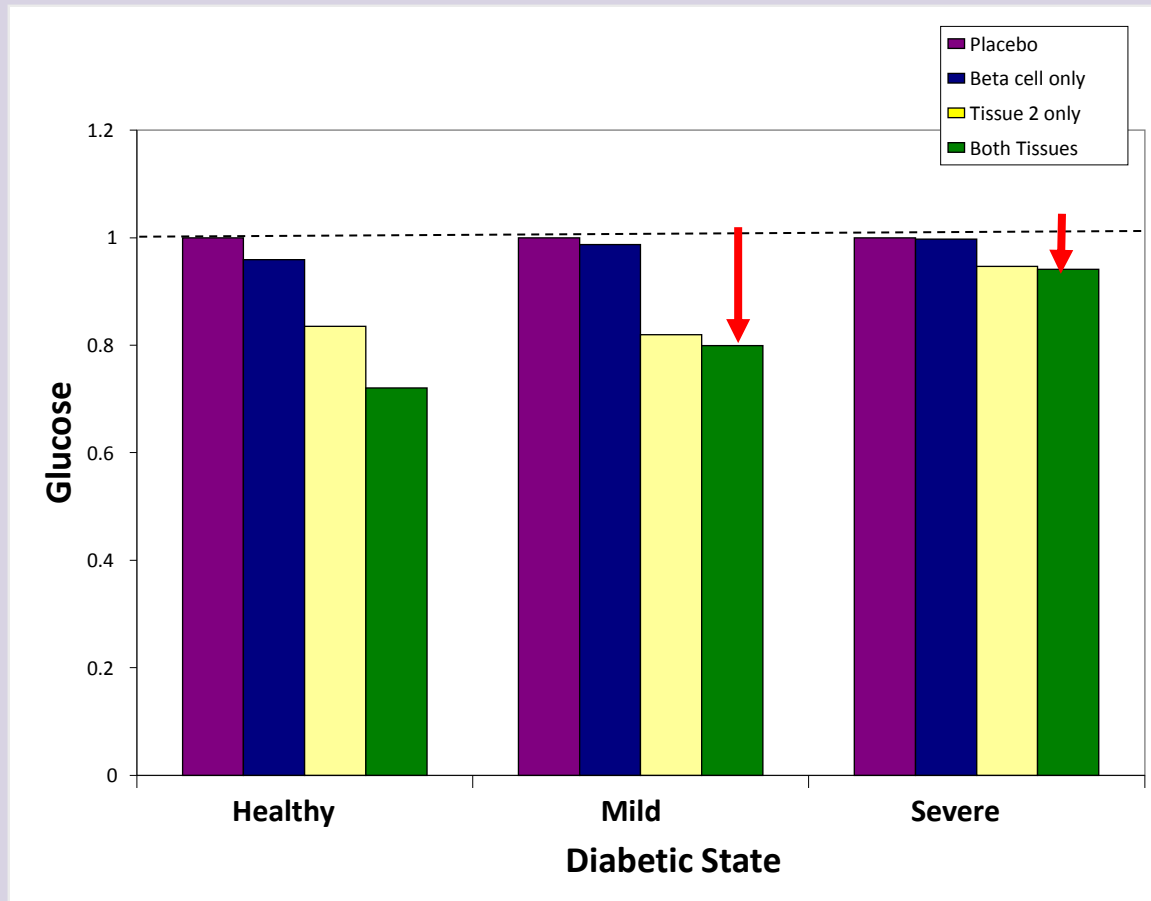
- Compound effect in the beta cell alone increased or maintained plasma insulin
- Compound effect in another tissue alone reduced plasma insulin
- The combination of these effects resulted in lower plasma insulin in diabetic VPs



Simulations in multiple VPs revealed that efficacy was also dependent on patient phenotype and pathophysiology.



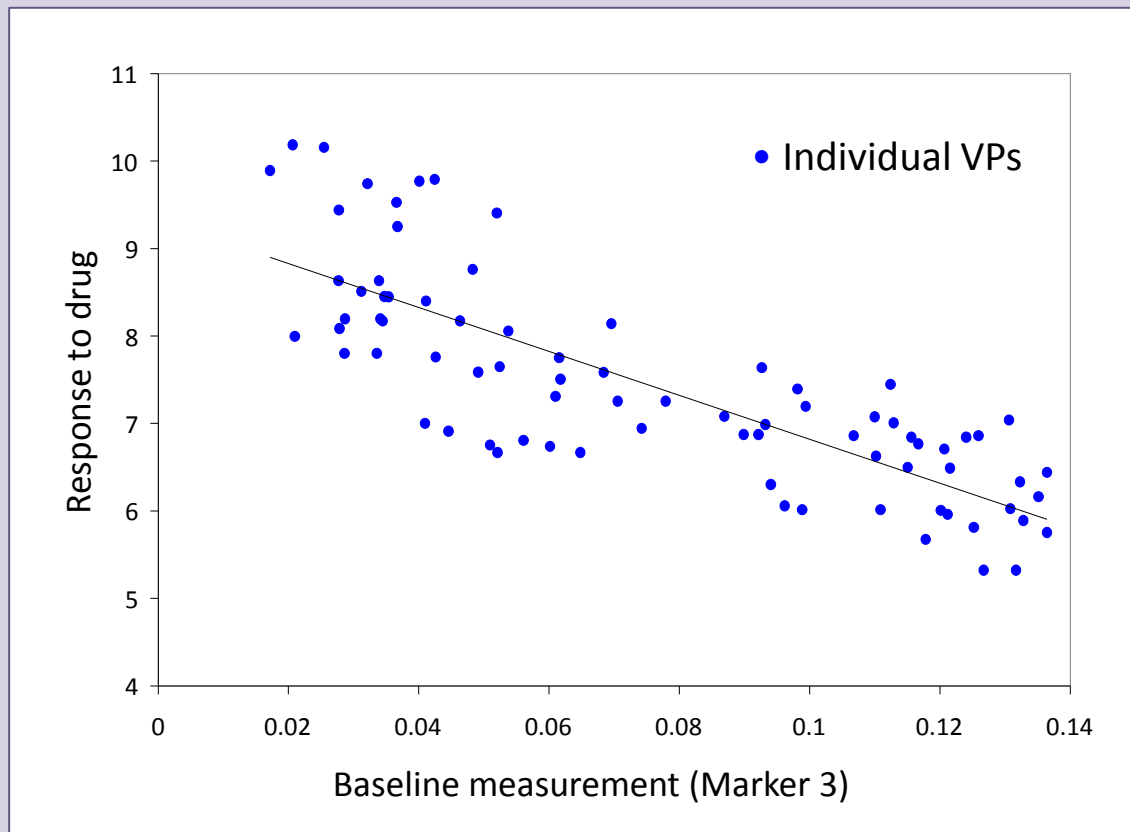
- Compound was less efficacious as diabetes severity increased
- PhysioPD research suggested this was due to reduced insulin secretory capacity



PhysioPD research identified a potential mechanistic biomarker distinguishing high responders from low responders.



Simulated biomarker 3 relationship to response



Marker	Correlation	P-Value
Marker 2	-0.107	0.4323
Marker 3	0.548	0.0083
Marker 4	0.004	0.9739
Marker 6	0.392	0.0026
Marker 7	-0.058	0.6872
Marker 8	0.254	0.0587

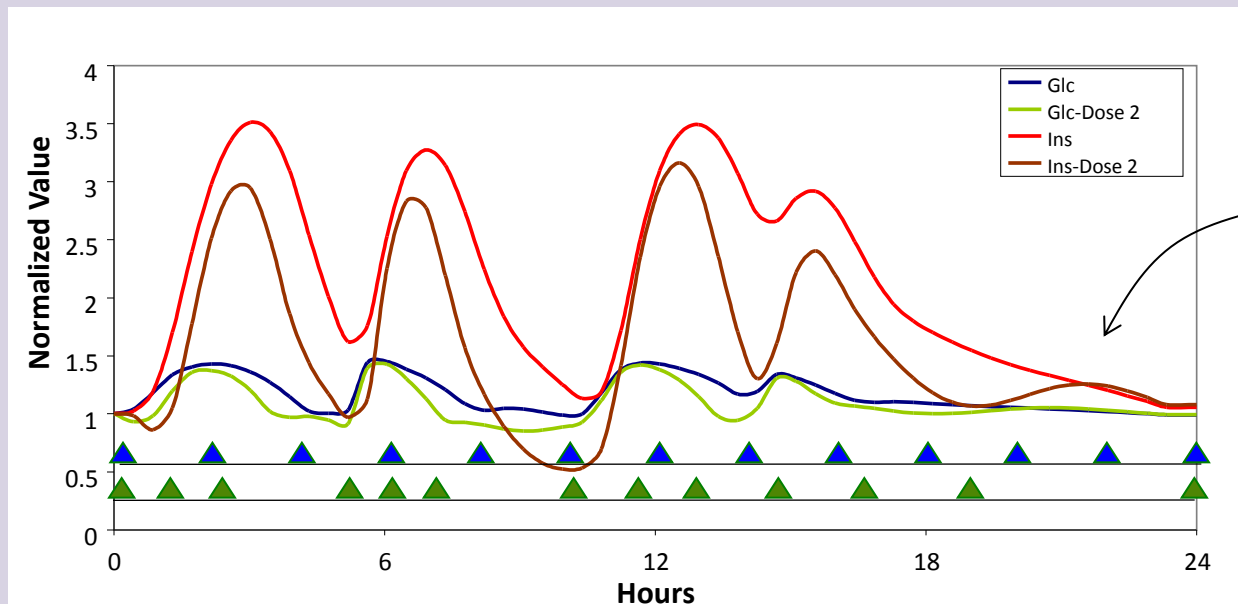


# PhysioPD research identified improvements for the proposed clinical trial design.



- Dose times relative to meals were optimized to increase sampling when treatment effect was greatest.
- Nighttime sampling was reduced without impacting trial predictive power.

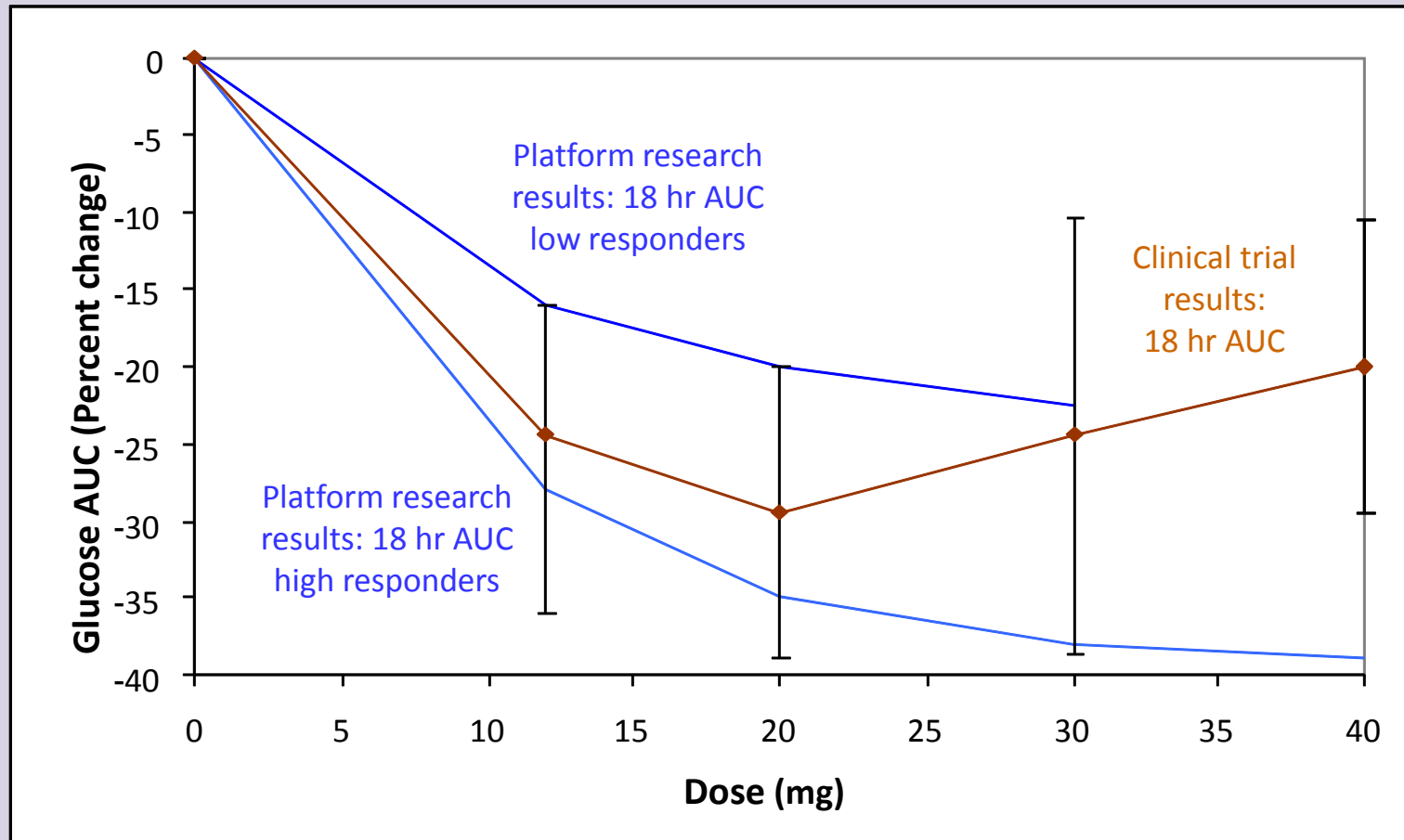
Glucose and insulin time course



Fewer samples  
needed at night

▲ Planned collection  
▲ Revised collection

PhysioPD research resulted in the design of a successful first in human clinical trial.



## Case Study Conclusions

- PhysioPD research gave critical mechanistic insight and guidance that optimized the clinical trial design and accelerated compound development
  - Aided interpretation of preclinical pharmacodynamic data
  - Identified responder and non-responder characteristics to guide patient inclusion criteria
  - Identified potential efficacy biomarker
  - Optimized sampling frequency to maximize opportunity to demonstrate treatment effect

## Summary and Conclusions

- PhysioPD research makes more complete use of existing data and biological knowledge, creates a bridge from mechanisms to outcomes, and facilitates:
  - Improved clarity and quantitative understanding of existing information
  - Efficient hypothesis generation and testing
  - Experimental designs that resolve key uncertainties and address variability
- By focusing on improving decisions, PhysioPD research has successfully impacted drug development in many, diverse therapeutic indications

***THANK YOU!***