



Systems Engineering Perspective of Human Metabolism

A Multi-scale Model for Disease Analysis Modeling Metabolic Health and Disease

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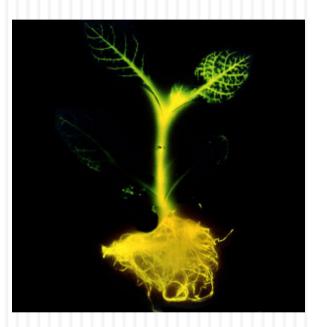
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Genotype to Phenotype

- □ Genome
- Transcriptome
- □ Proteome
- Metabalome
- Phenotype

Presence of genome does not ensure a phenotype It requires a specific state in the hierarchical chain.

Central Dogma of Biology



Luciferase Gene

Luciferase gene decoded

Catalyzed by Luciferase Enzyme



Phosphorous release using ATP

Firefly Glows Transgenic Plant made to Glow Gene

RNA

Protein

Metabolic Reaction

Phenotype

Descriptional Science

- Historically Biology has been a descriptional science
- Molecular Biology has given molecular level description of the Hierarchical state.
- □ Genetic, signaling/protein and metabolic networks are the result of reductionism of Molecular Biology.
- Bioinformatics has added more information to this approach.
- Resulting in a highly interconnected network!

Quantification of Systems

- Engineering systems are quantified to a level that they are designed, optimized and optimally operated.
- □ Biology is evolving to be a quantitative science.
- Principles of system science can be applied to biology: Systems Biology

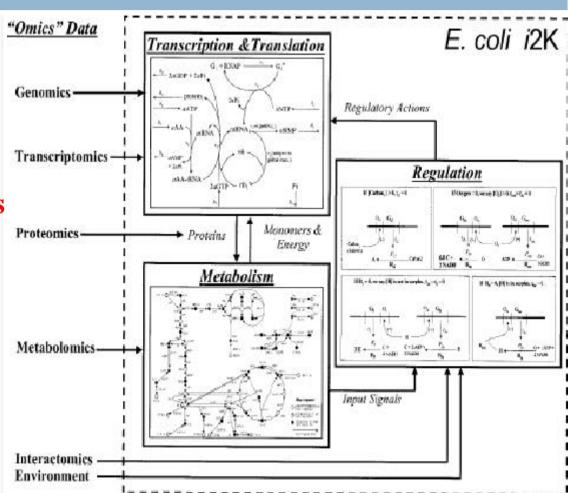
Bottom-up Design of a Complex System



- 1250 computers
- Hundreds of feedback loops
- **Millions of components**
- Design Manual Available
- Only take off requires manual effort

Design in Nature: Top-Down Approach (Escherichia coli)

- About 4400 genes
- Connectivity between genes,
 mRNA, proteins & metabolites
- Thousands of feedback loops
- No design principles available
- No computation control & sensing achieved through interactions of biomolecules



Complexity in Engineered and Natural Systems

- Non-linear dynamics
- Multiple feedback loops
- Multiple interactions
- □ Cascade structures
- □ Feed forward loops
- Interactions between modules
- □ Timescale separation

Resulting in a Complex system

Complexity in Engineered and Natural Systems

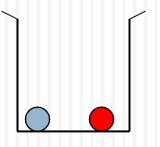
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Resulting in a Complex system

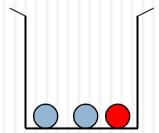
Example of a "Simple" Complex System

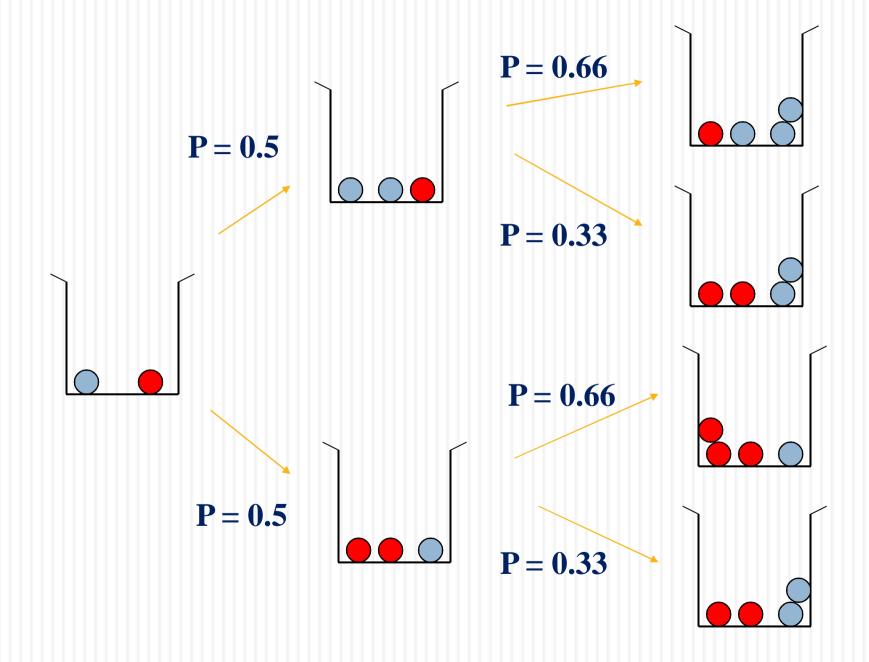
Experiment:

- 1. One **Red** and One **Green** Ball in a Box
- 2. Choose a ball randomly
- 3. Add and replace the same color ball into the box
- 4. What is the fraction of red balls after 50,000 steps

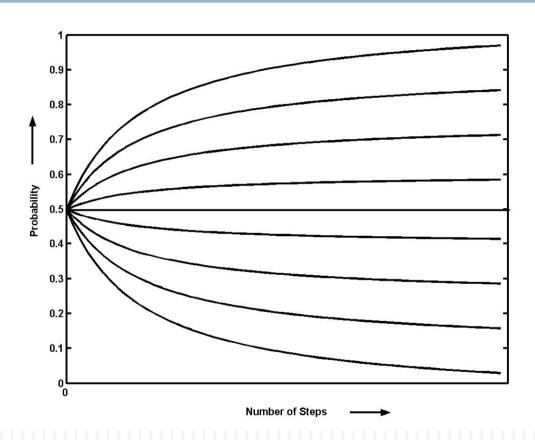


Picked Orange ball Replaced the same and added another of same colour





Possible Outcomes in the Experiment



Multiple outcomes possible depending on initial few steps

Properties of Complexity

- Multiple Steady States
- Depends on History (Initial Condition)
- □ Parametric Sensitivity
- Bifurcation
- □ Robustness
- Adaptability
- Evolvability

Reconstructing Cellular Functions

Palsson BO, Nature Biotech 18:1147 (2000

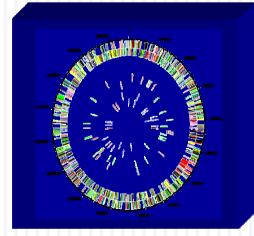
Reductionistic Approach:

Components Biology

HT analytical chemistry genomics transcriptomics proteomics

20th Century Biology





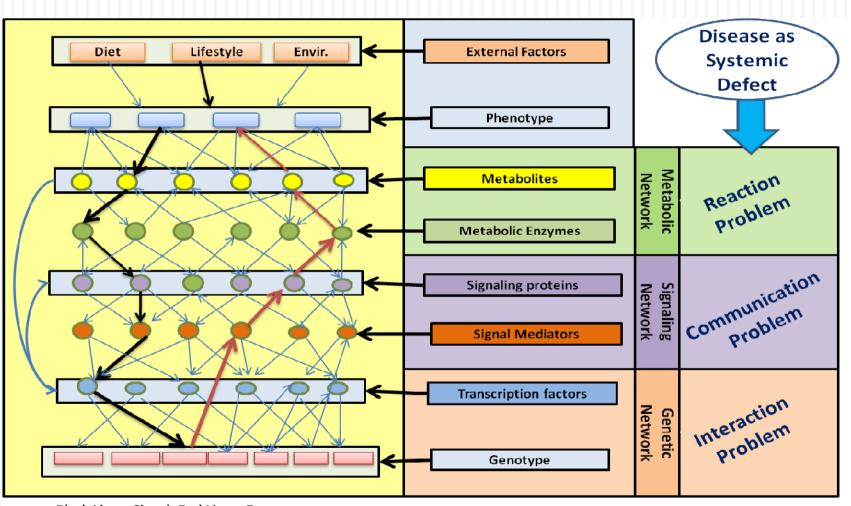
Integrative Approach:

Systems Biology

- 1. Components
- 2. Reconstruction
- 3. Modeling/simulation
- 4. Feedback analysis

21st Century Biology

Disease as Systemic Defect



Black Lines- Signal, Red Lines- Response

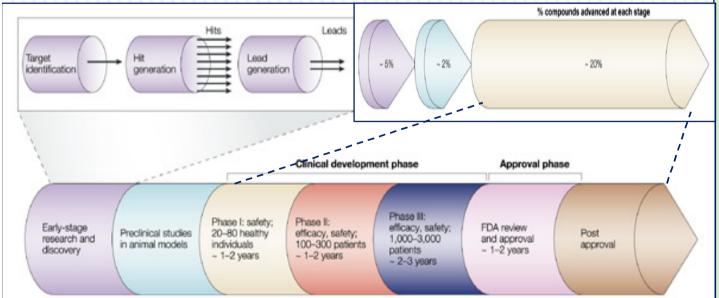
(Pramod and Venkatesh, 2013)

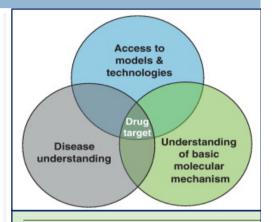
Challenges in Modeling Disease States

- Incomplete interaction map
- Incomplete Parametric space
- Ambiguity in data
- Difficult to validate model
- Need for a composite clinical data
- Validation is in the overall physiological response than quantitative
- Incomplete and ambiguous disease characterization

Systems Biology in Drug Discovery

- ➤ Basic research and Target Identification costs about 20-25 % of total drug development cost
- ➤ Only 2-5 Targets per 1000 targets reach the final stage of drug development
- ➤ Huge amount of work, time and money consumed





Systems level insights

Hypothesis generation

Model testing

Target identification/validation

Compound validation

Lead optimization/SAR

Disease indication/ trial design

Biochemical Networks-Structure and Dynamics

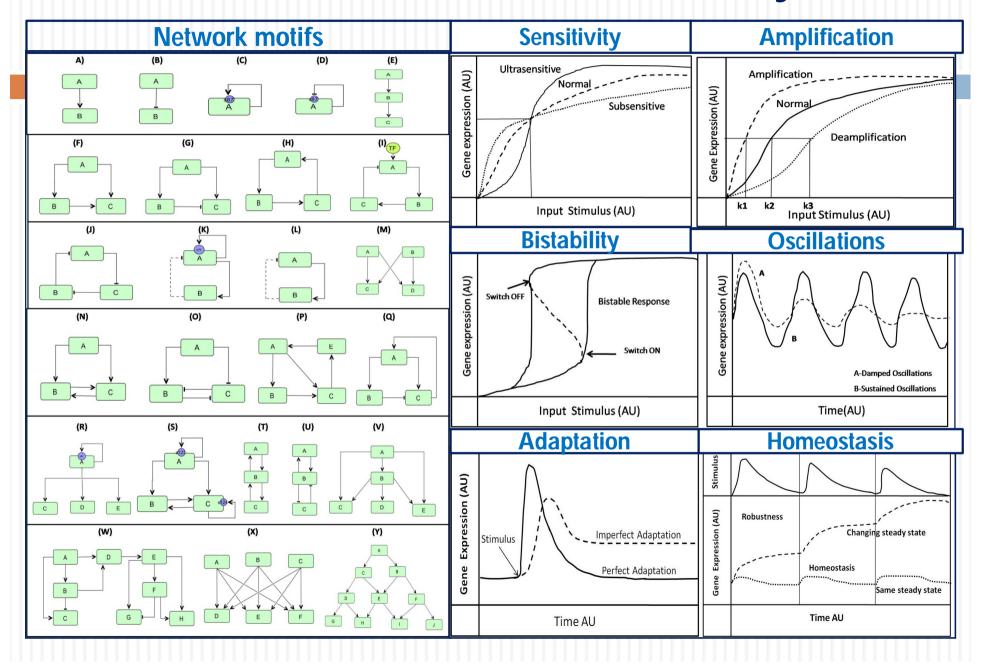
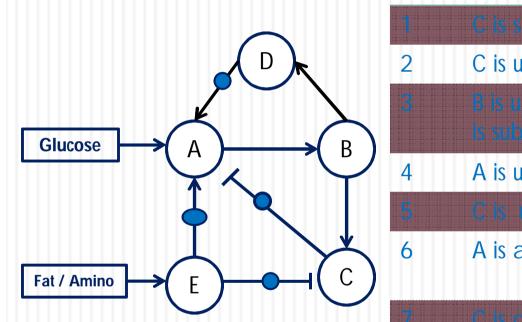


Illustration of Metabolic Regulatory Dynamics



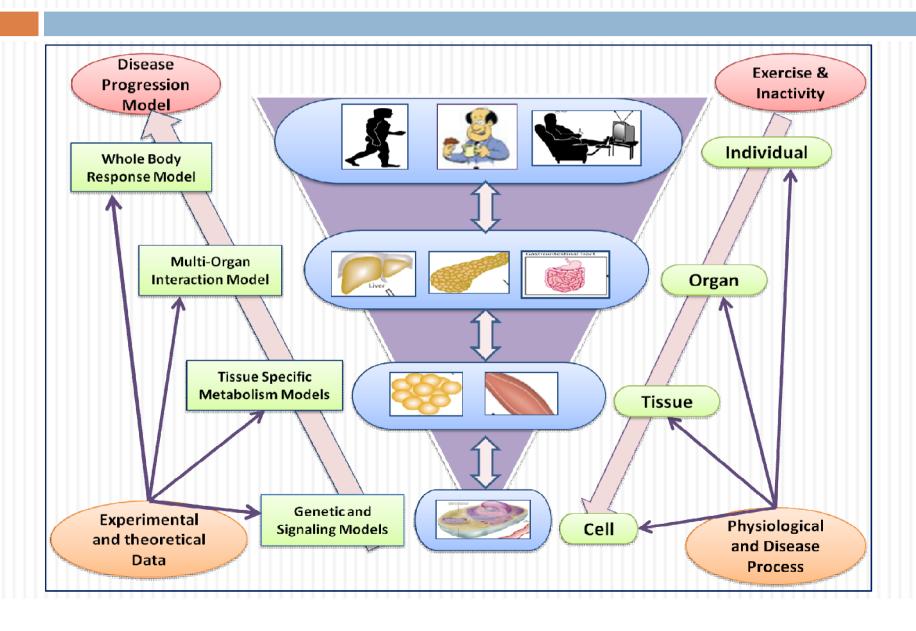
Case	Description	Response
1		
2	C is ultrasensitive to B	Homeostatic
3	B is ultrasensitive to A and A is sub sensitive to C	Oscillation
4	A is ultrasensitive to D and C	Robust
.[3]	C is ultrasensitive to E	Bistable
6	A is amplified by D and E	Irreversible Bistable
7	C is deamplified by E	Damped Oscillation

Sensitivity and amplification are governed by the regulatory mechanisms (Signaling and trancription) and mutations, allosteric interactions, conformational changes etc.

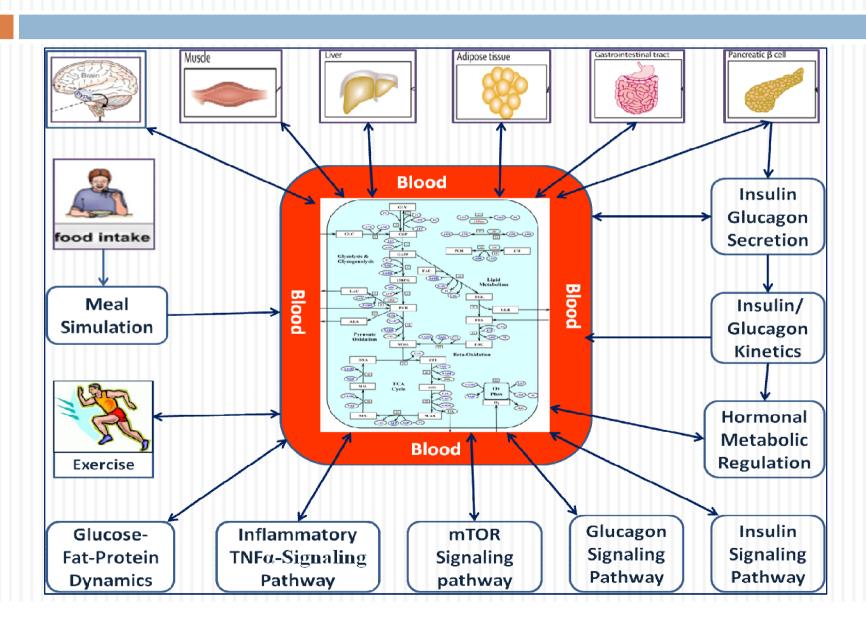
Introduction – modeling whole body metabolism

- A Multi-scale Composite model for whole-body Human Metabolism incorporating tissue specific metabolism and regulation
- Incorporates Carbohydrate, proteins and Lipid metabolism including their regulation through signaling network
- Goal towards analyzing effect of lifestyle such as diet and exercise on metabolism
- Deciphering design principles and control structures for healthy and disease states
- Analysis of disease states such as metabolic syndrome and cancer
- Sensitivity analysis towards rate limiting steps for therapeutics
- System level dynamic property evaluation of homeostatic variables

Modeling Strategy

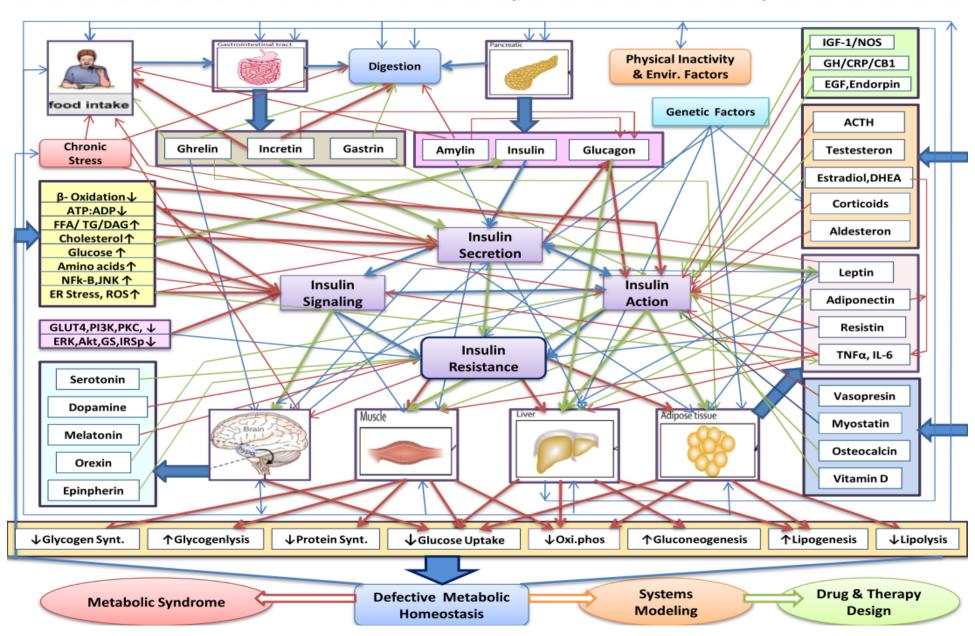


Composite Model



Insulin Resistance-Problem Complexity

Blue lines-Process Flow, Green lines-Positive Interactions, Red lines-Negative Interactions, Bold lines-Pathways and models available



Hormonal Metabolic Regulation

Insulin (Anabolic)

Glucagon (Catabolic)

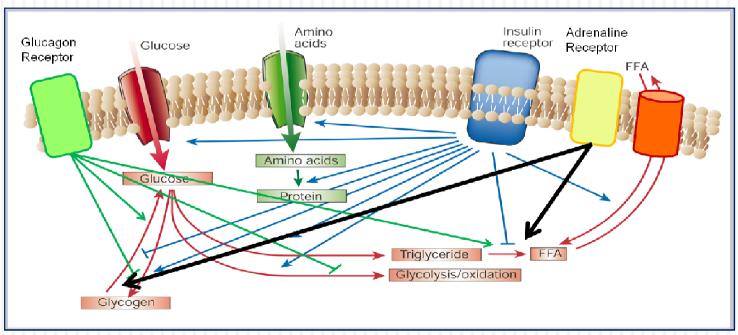
Adrenaline (Neural)

- Glucose uptake
- Glycolysis
- Glycogen synthesis
- Protein synthesis
- > Fat Synthesis

- Glycogenlysis
- Gluconeogenesis
- Fat breakdown
- Proteolysis
- Ketogenesis

- Glycogen breakdown
- > Fat breakdown
- Increase in Blood flow

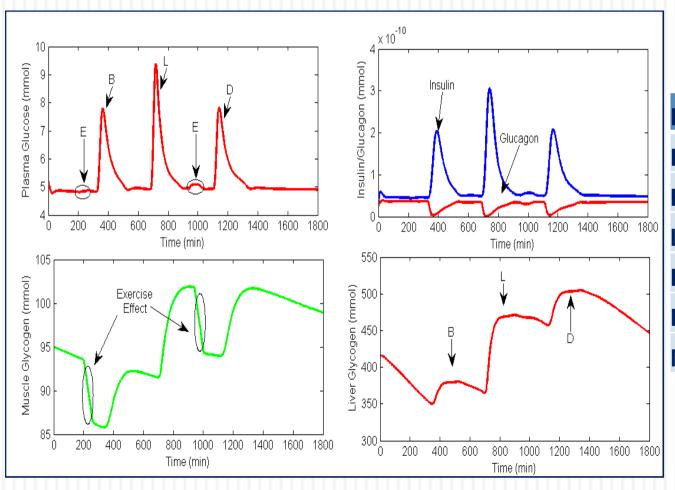
(During Higher work rate and Exercise)



(Modified and adopted from Saltiel and Kahn 2001)

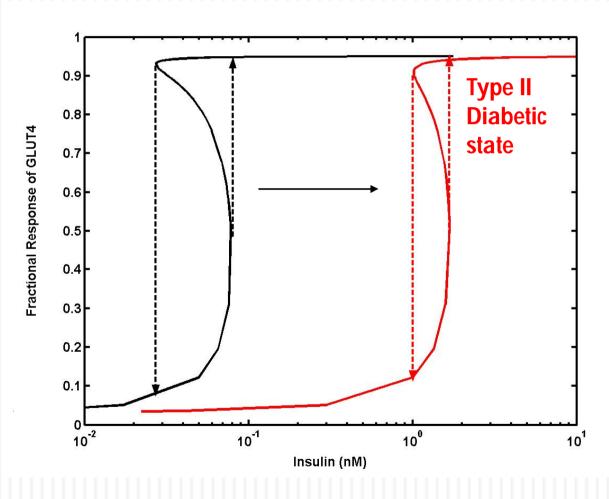
Typical Lifestyle Simulation

B-Breakfast, L-Lunch, D-Dinner, E-Exercise



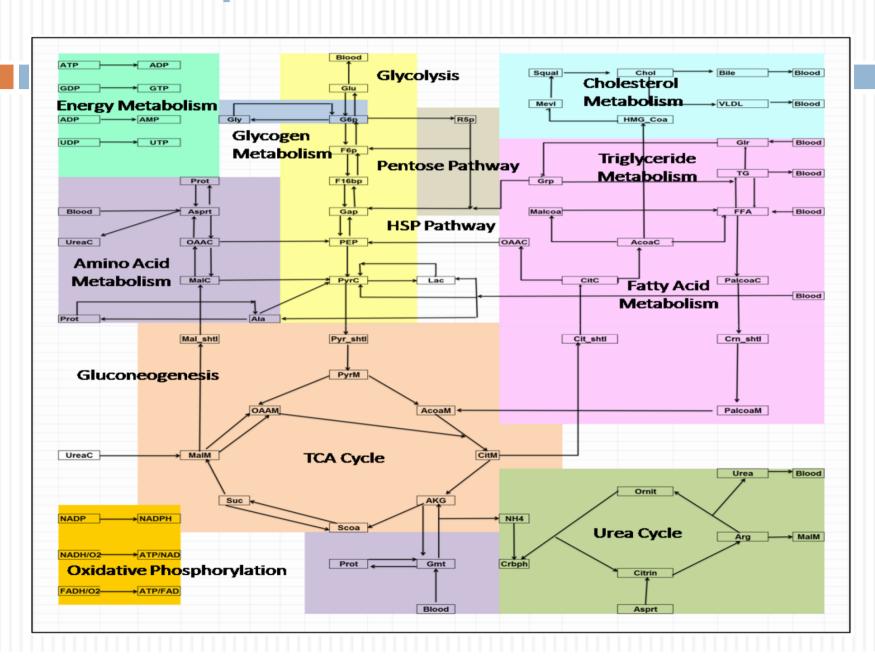
Input	Qty	Time
Exercise-1	1 hr	6 AM
Breakfast	50 g	8 AM
Lunch	100 g	2 PM
Exercise-2	1hr	6. PM
Dinner	50g	9 PM
Night-Rest	11 hr	

Bistability in Insulin Signaling Pathway Type-II Diabetic State – Effect of PTP

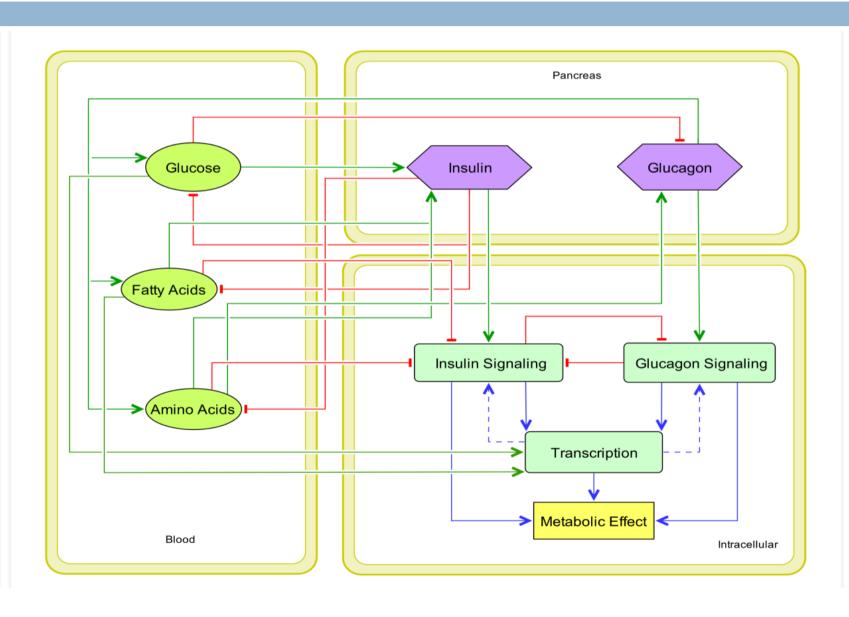


Giri, Mutalik, Venkatesh, BMC Theoretical Biology & Medical Modeling

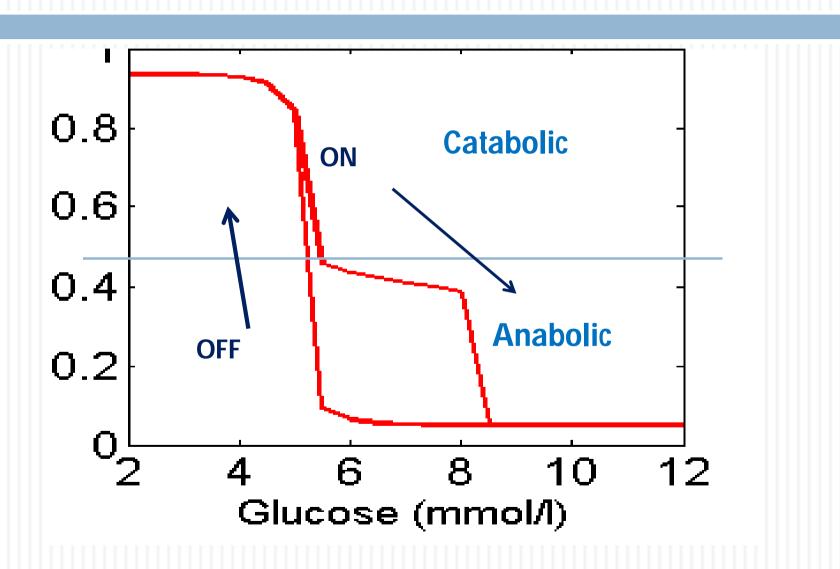
Hepatic Metabolic Network



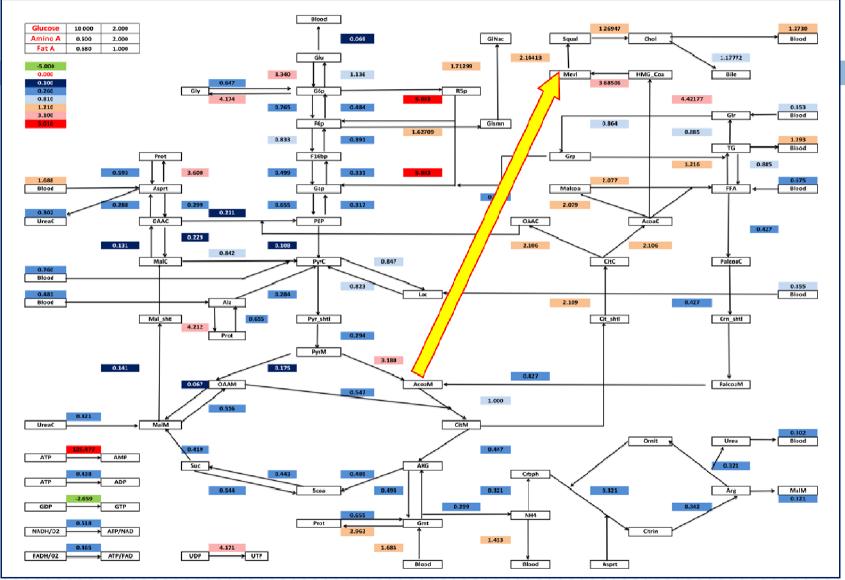
Linking metabolism and signaling



Steady state response for Glucose and Glucagon

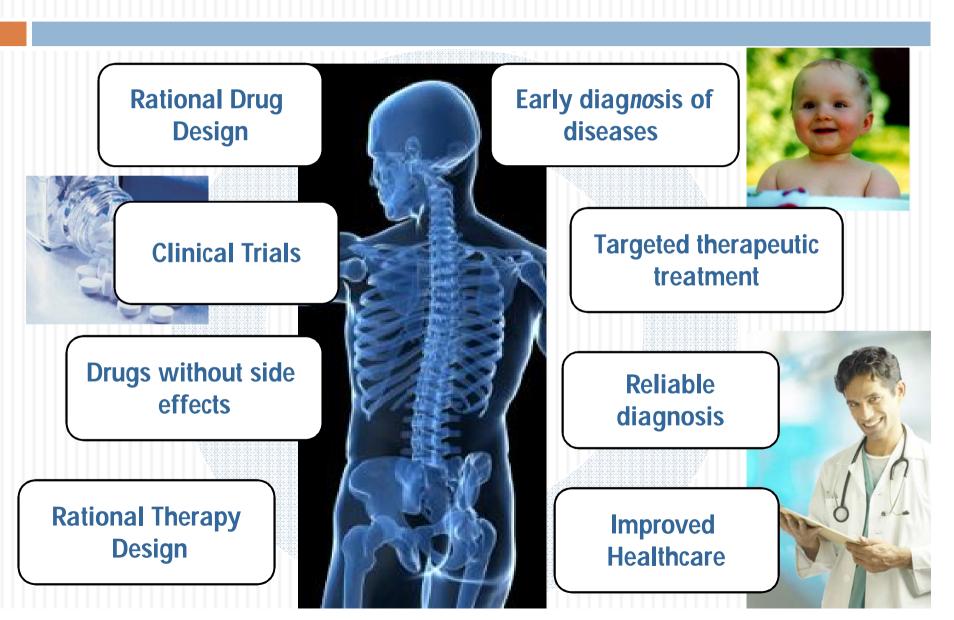


High CH and Protein, normal fat



The yellow arrow shows the diversion of the metabolic flux towards the lipogenesis.

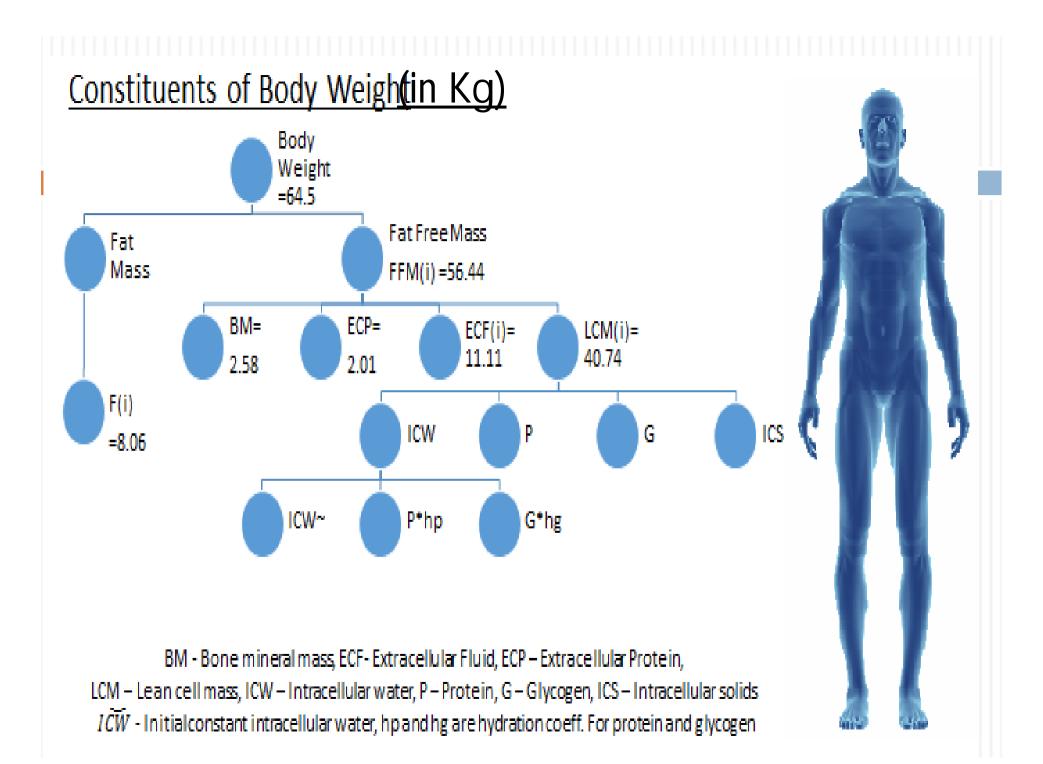
Significance of Disease Modeling



How Diet Affects Human Body Weight



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Venkatesh
Department of Chemical
Engineering



Background

Obesity is being an epidemic in the current world. Obesity causes many fatal diseases. The prevalence of obesity is alarmingly high, and the proportion of adults who are overweight and obese in the developed country like US continues to increase.

The complexity of interactions among dietary and genetic factors is becoming more appreciated, and much research has been undertaken in recent years to probe various interactions between diets and its effect, which has allowed for a greater understanding of the intricacy of this field of research. Body weight depends upon dietary intake, surrounding and physical activity. Body weight can be controlled by changing our dietary intake and daily life style.

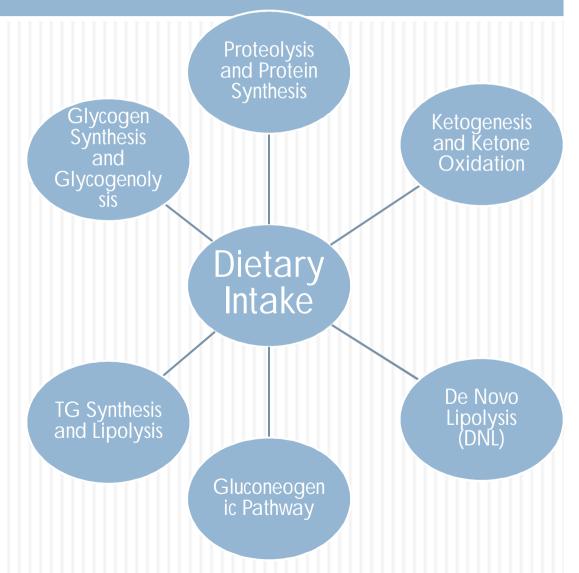


Dietary Intake

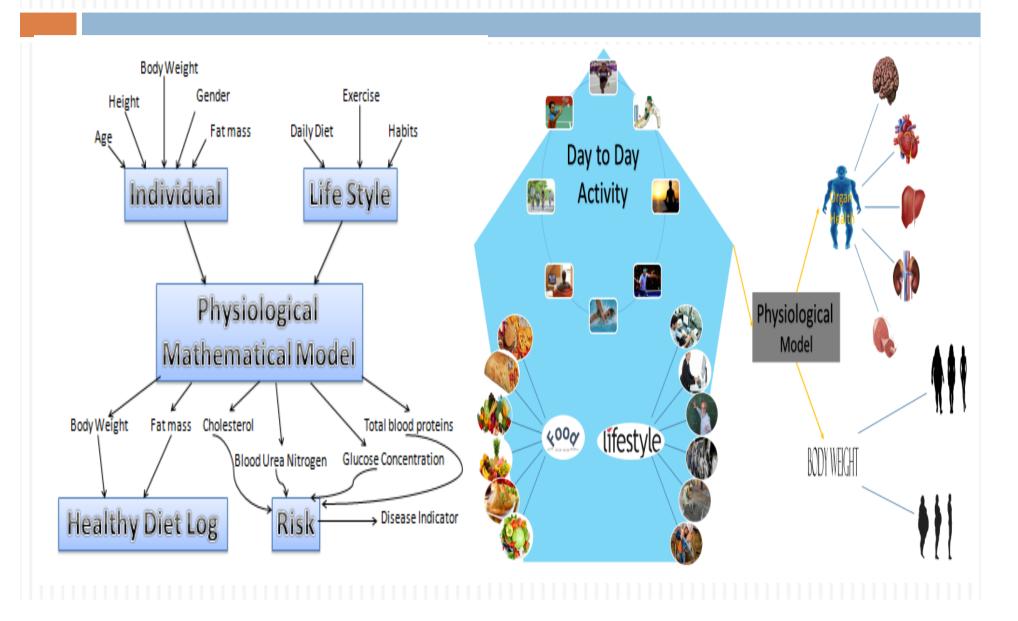
Carbohydrate, fat and protein goes for different metabolic pathways.

Glucose can be made by fat and protein as well for energy requirement, using the gluconeogenic pathway.

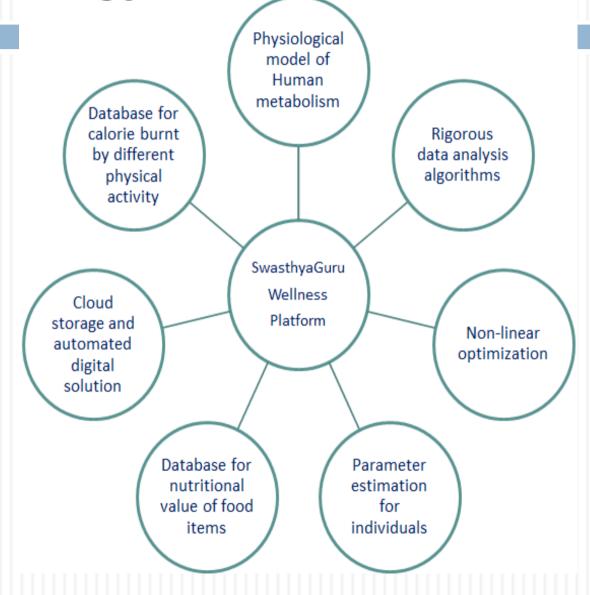
Carbohydrate can be converted to fats using De Novo Lipogenesis metabolic pathway.

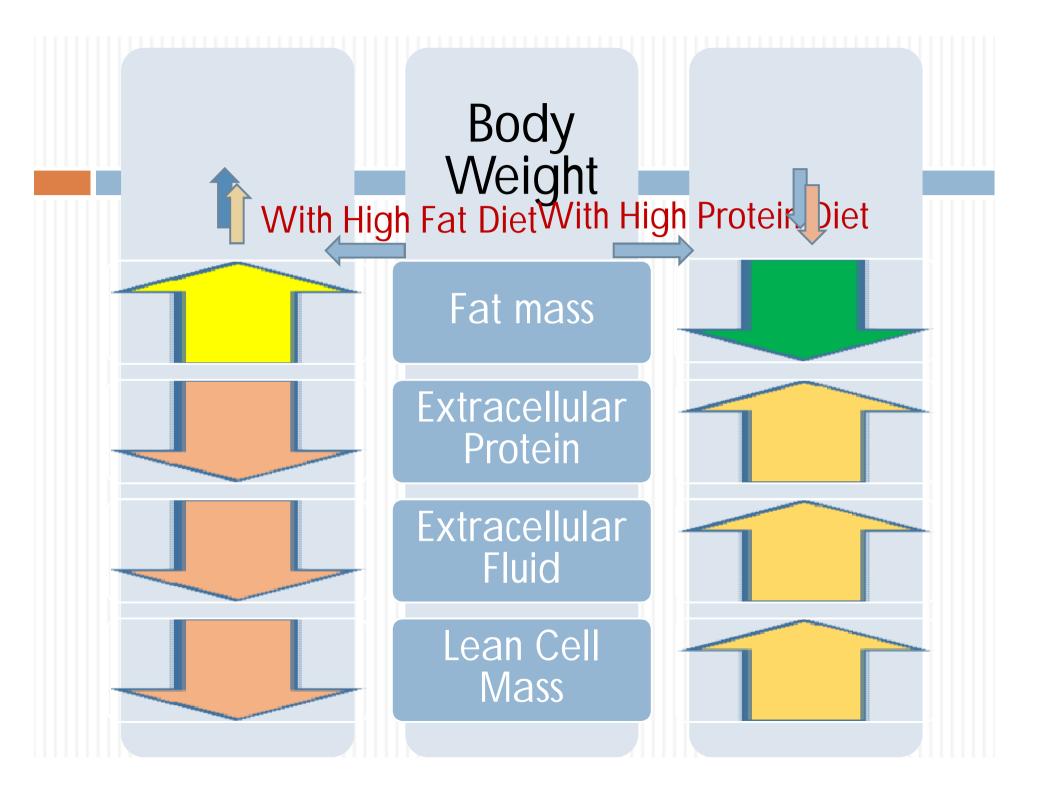


Input output flow to the model



Technology used

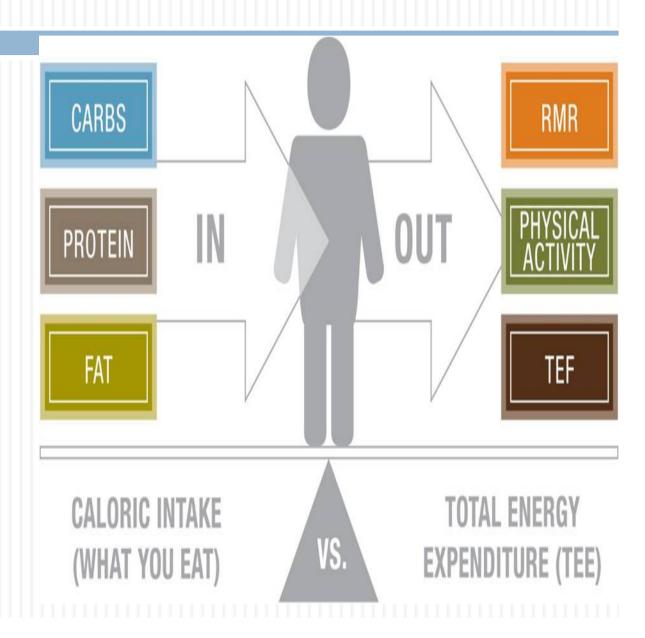




Energy Intake and Expenditure

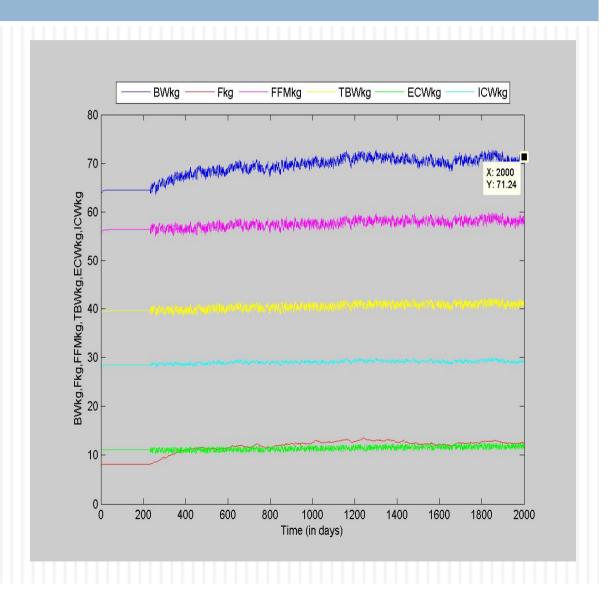
Thermic effect of food (TEF), also known as specific dynamic action (SDA) of a food or dietary induced thermogenesis (DIT), is the amount of energy expenditure above the resting metabolic rate due to the cost of processing food for use and storage.

Resting metabolic rate (RMR) is the minimum number of calories your body needs at rest to fuel its metabolic activity, for example to maintain functions such as heart beat, breathing and temperature.



Fluctuations in body weight while diet remains in between $\pm 50\%$ of normal diet

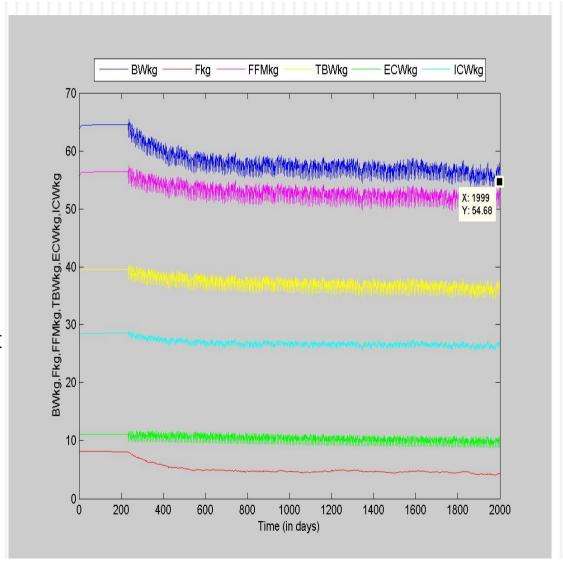
Body weight increases by consuming random diet.
Random diet is not advisable to maintain the body weight.
Carbohydrate oxidation is observed to be increased.
While increased fat oxidation helps in reducing/maintaining the body weight. We can increase fat oxidation by fasting a day in a week.



Body weight with fast a day in a week with diet remains between $\pm 50\%$ of normal diet

I simulated the model with zero calorie on the day of fast. Body weight These results are for sedentary person like office worker.

Body weight can be reduced by following this schedule. Fat oxidation increases which reduces the fat mass.



Ode used for TC, TG, Nitrogen, Urea

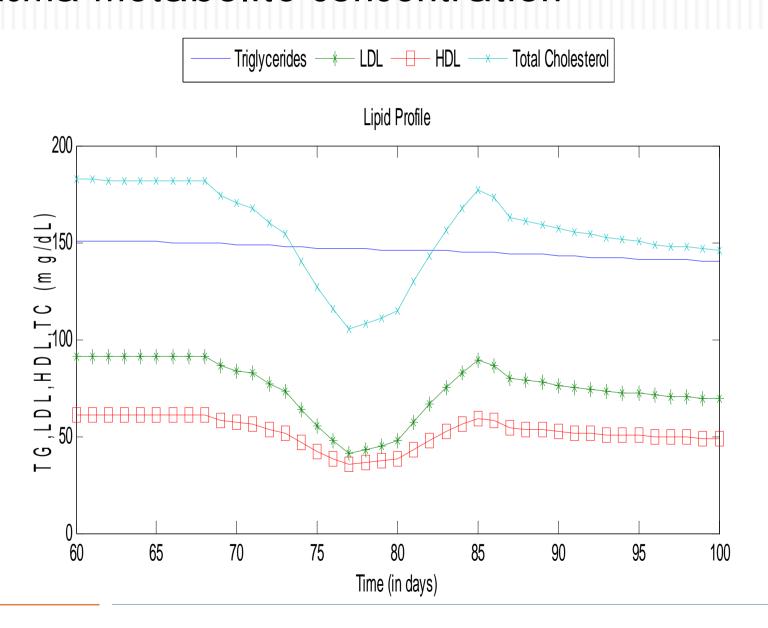
$$\bullet \frac{D[Chol]}{dt} = \eta_{chol} * k_{Chol_synth} - k_{util} * Chol + k_{abs} * Chol_{food}$$

$$\bullet \frac{D[TG]}{dt} = k_{TG_{synth}} * \eta_{TG} - k_{brk_{dwn}} * TG$$

$$\bullet \frac{D[N]}{dt} = k_{basal} - k_{des_N} * N - k_{net} * Prot_{diet}$$

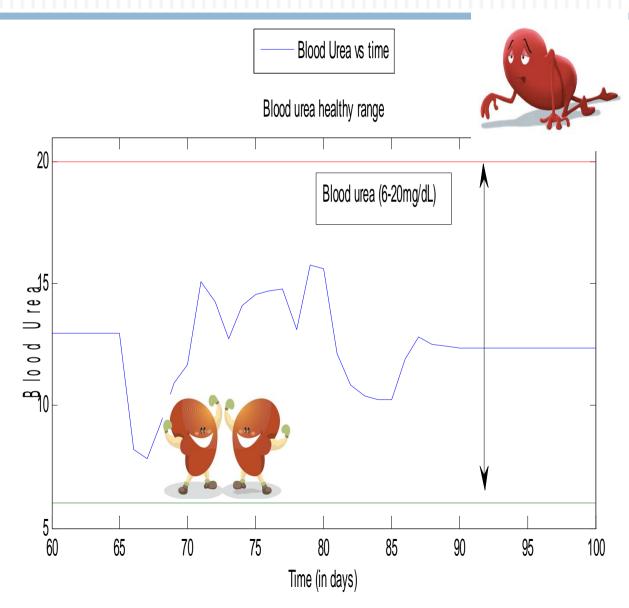
$$\bullet \frac{D[U]}{dt} = k_{synth} * \eta_{u} - k_{brk} * U$$

Plasma metabolite concentration

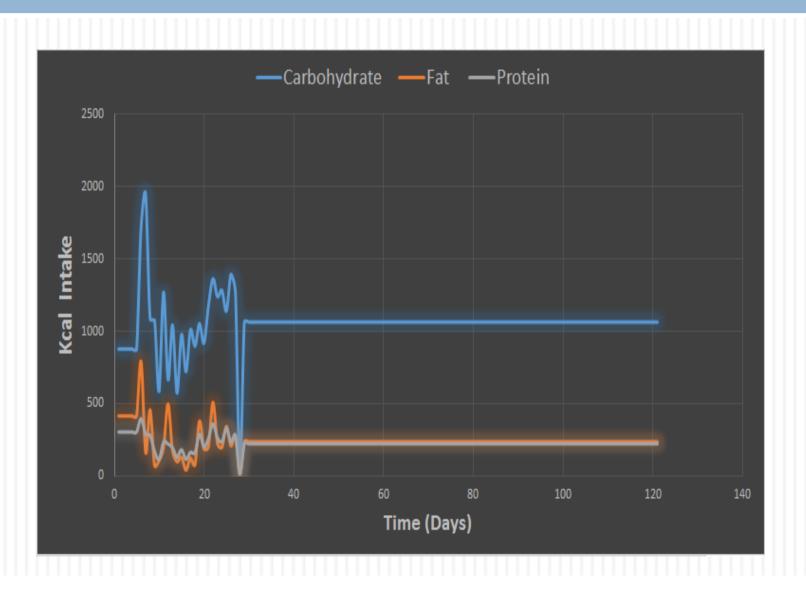


Predicting health of Kidney

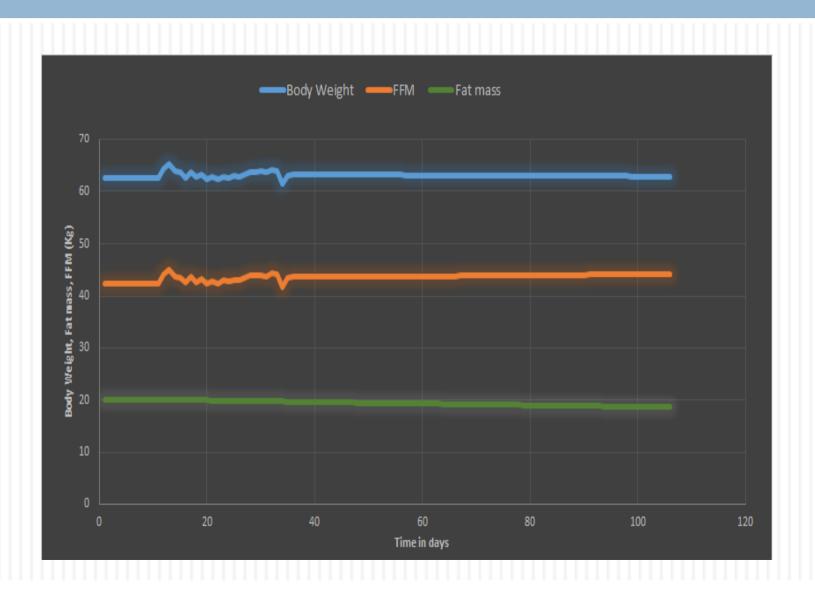
- High BUN
 means kidney
 is injured
 which might be
 caused by
 diabetes or
 high blood
 pressure.
- Low BUN represents low protein intake



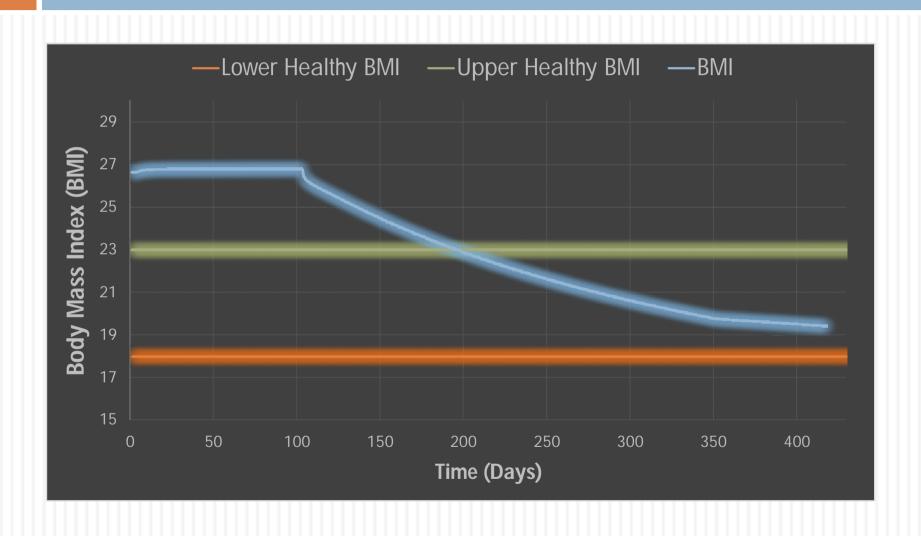
Kcal Intake



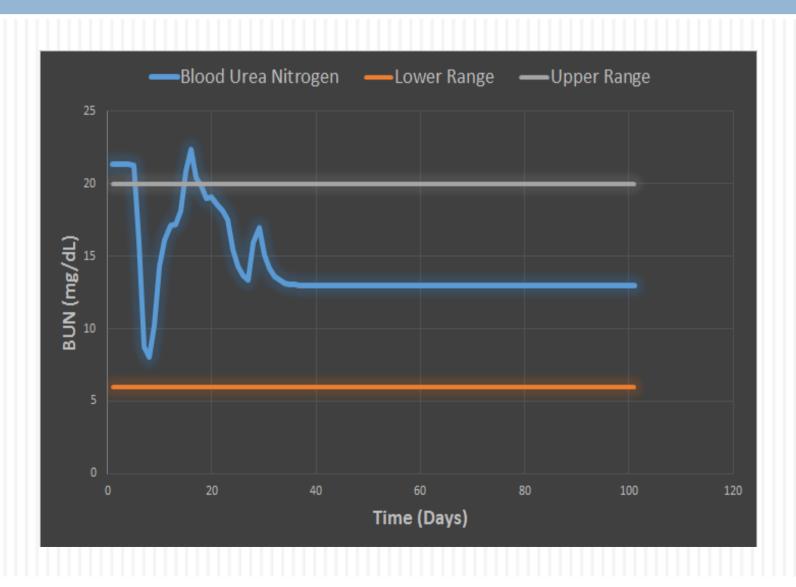
Output: - Body weight dynamics



Body mass index

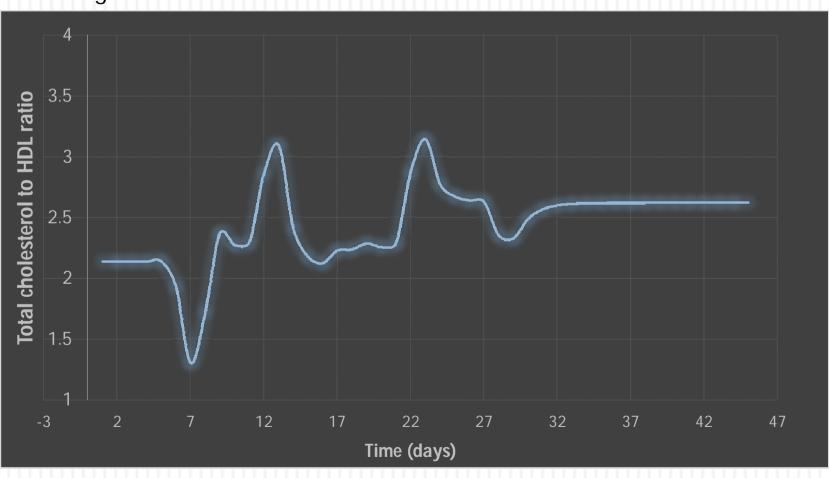


Blood urea nitrogen

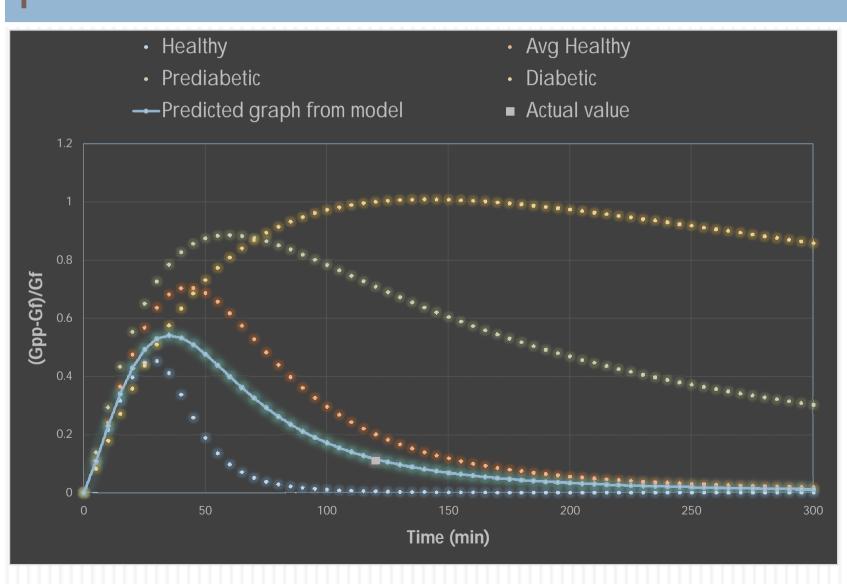


Risk for diabetes

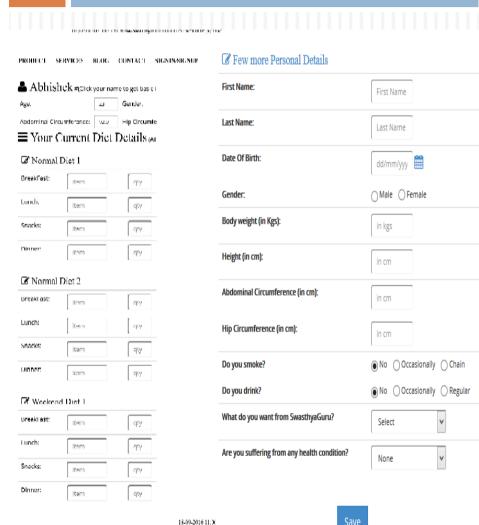
Total to HDL ratio should be in between 3.5 to 1. A ratio greater than 3.5 exhibits risk of diabetes



Glucose dynamics for diabetes prediction



Input to model and predicted output





67

71.2

104

159

14

7.2

30-70

<130

25-160

140-250

6.0-23.0

6.0-8.0

52

87

110.2

161

12.9

7.6

Optimized Suggested Plan for ideal body weight

Recommended Plan:
Desired final weight

0.5 Kg loss per month
66 Kg (BMI: 21.5)

With Diet only Plan: 2020 (241 Kcal reduction per day)

HDL

LDL

Triglyceride

Total Cholesterol

Blood Urea Nitrogen

Total Blood Protein

With exercise only Plan 35 minutes of cycling per day with original 2261 Kcal intake

PS: Above plan can be modified to suite your choice (combination of diet and exercise)

Unveiling

Predict your health risk with SwasthyaGuru

Login your current lifestyle to know your future health risk

SwasthyaGuru will help you manage your lifestyle for the better

Futuristic tool for personalized diet and activity managements

Our capabilities

- System level human metabolic model to predict future physiological parameters based on current lifestyle inputs specifically for Indians.
- Adaptive weight-loss/gain management

 Maintain and sustain weight with ideal fat mass/lean
 muscle mass through optimized diet and exercise routine
- Food and physical activity diary
 Day to day diet and physical activity log to help balance calorie intake and expenditure

Our capabilities

- Micro-nutrient balanced diet (Calcium, iron, vitamins)
 Minimize deficiency in micronutrient for better health and active lifestyle
- Predict and control physiological parameters (Blood glucose, cholesterol, urea etc.) towards reduced risk of diabetes, cardiovascular, PCOS
- Medical Repository

Store your medical history (Blood profile, X-ray, ECG, Pathlab reports) at one place for better future health management

Unique Selling Point

- Knowledge based human physiological model for predicting health parameters
- Individualized recipe for weight management through an automatic optimized solution platform based on a database of Indian foods
- Personalized plans based on the choice of the individual
- Disease management solutions for lifestyle diseases such as diabetes, heart disease, PCOS etc.
- Predict and maintain blood plasma metabolite profiles for better organ health
- Micronutrient balanced diet
- Platform for infants and children (a solution for the whole family)

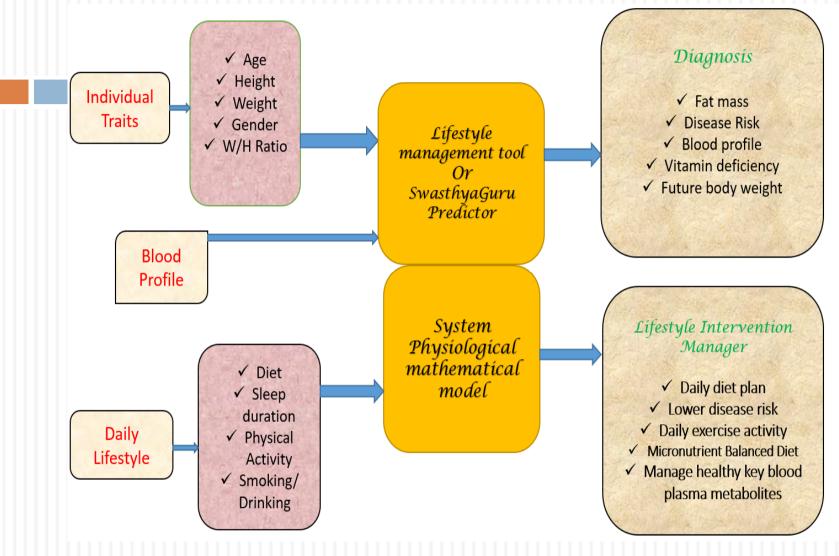
Capabilities of the Platform

Micro-Scale Model:

- 1. Quantification of metabolism and signaling for individual organ
- 2. Effect of diet and lifestyle on intracellular pathways
- 3. Analyzing disease progression and diagnostics
- 4. Effect of active metabolite/compound on whole body physiology

Macro-Scale Model:

- 1. Prediction of body weight and fat percentage
- 2. Individualize the model for person specific through parameter estimation
- 3. Disease risk analysis based on lifestyle inputs
- 4. Optimized diet for both macro and micro nutrients and physical activity chart
- 5. Lifestyle intervention for disease management
- 6. Medical Data repository for individuals



Thank You

Acknowledgements

- 1. Dr. Pramod Somavanshi, currently PostDoc at Harvard Bioengineering Department
- 2. Mr. Abhishek Koshta, Mtech student
- 3. Mr. Jyot Atnani, Btech Student
- 4. Ms. Manaswita S, Btech Student

THANK YOU