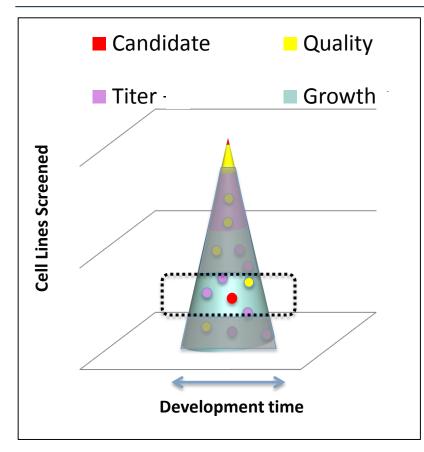
Cell Line qualifying attributes for prediction of good producers

Alessandro Mora

UMAss Lowell, Biopharmaceutical Summit

Lowell May 28, 2014

Cell line development aims to select the "best" candidate

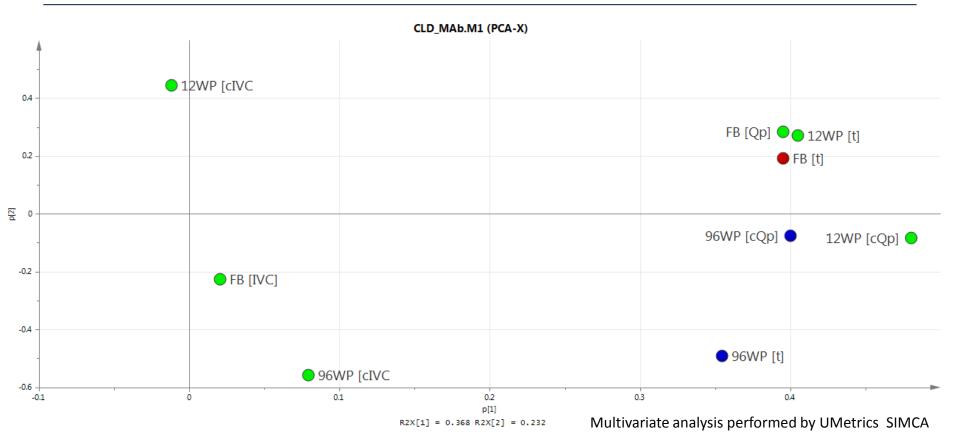


	Titer [t]	Growth [IVC]	Productivity [Qp] (titer/growth)
96WP	96WP[t]	96WP[IVC]	96WP[Qp]
12WP	12WP[t]	12WP[IVC]	12WP[Qp]
96DWP fed-batch [FB]	FB[t]	FB[IVC]	FB[Qp]

96WP[IVC] and 12WP[IVC] attributes were generated by corrected confluence analysis

Three different parental lines originated 320 sub-lines whose 9 attributes were screened throughout the development in order to **explore hypothetical patterns**

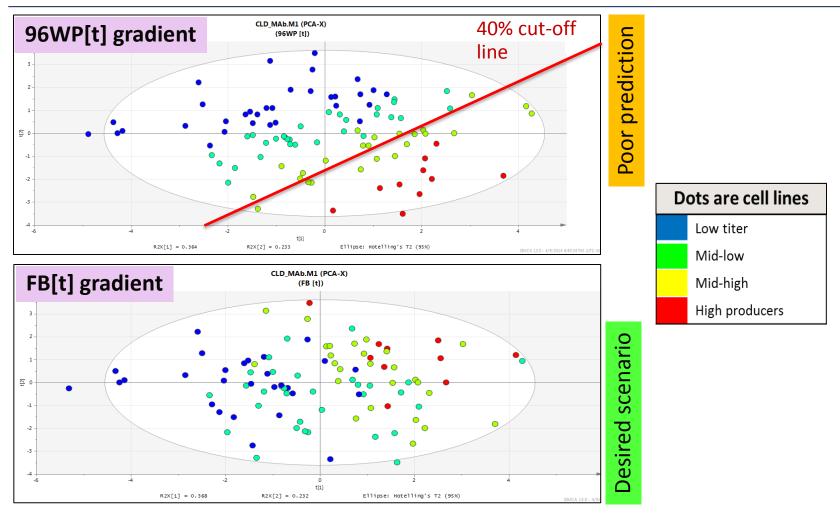
96WP[cQp] might be a better qualifying attribute than 96WP[t]



Multivariate Analysis explains relationships among all the attributes studied.

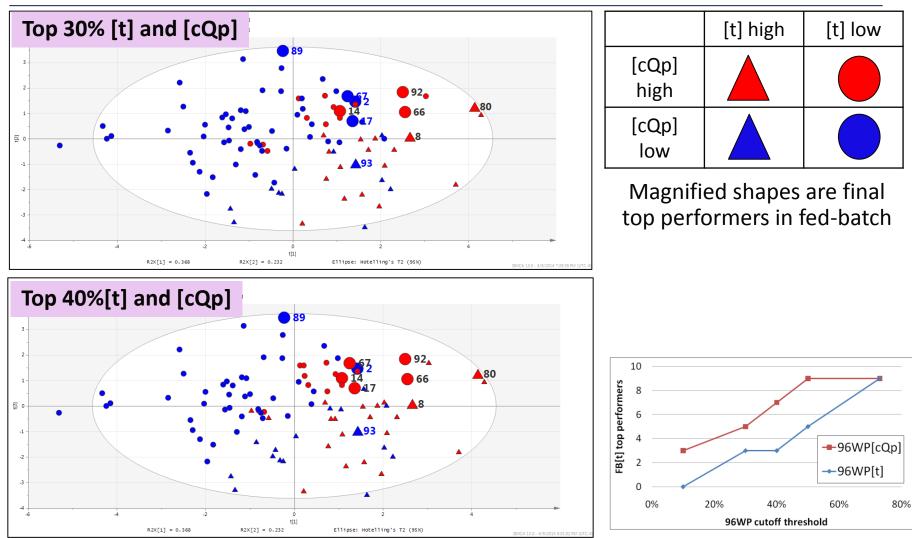
- Attributes closely located behave similar (positive covariance).
- Attributes far away from the origin **drive the** *"best cell line"* model.

Using static titer to predict final fed-batch titer



Final top performers in fed-batch are poorly predicted by static titer at 96-well plate. Vessel adaptation, cellular dynamics, mechanics and genetic events contribute to this.

Top 96WP[cQP] and 96WP[t] performers Vs. Top 10% FB[t]



96WP[cQp] is a powerful tool and captures more final candidates than 96WP[t].

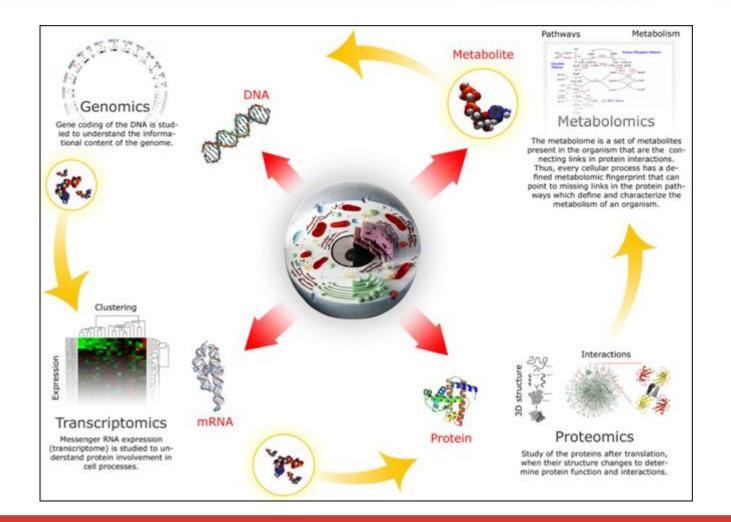
- Early stage titer in 96WP doesn't predict top FB producers.
- Productivity in 96WP shows a good predictive power combining existing 96WP[t] with new developed 96WP[IVC] attribute.
- 96WP[cQp] acts like an enhanced prediction tool and it might save about 20-40% initial workload in cell line development.



Metabolomics for CHO-Cell Production

Seo-Young Park Ph.D student, Chemical Engineering, Umass Lowell 05/30/2014

Overview

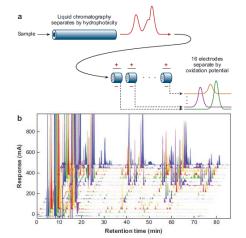


Genomics and **Proteomics** tell you what <u>might</u> happen, but **metabolomics** tells you what actually <u>did</u> happen!

The conceptual approach in metabolomics



Sample Collection

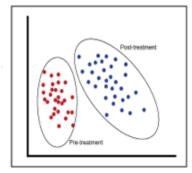




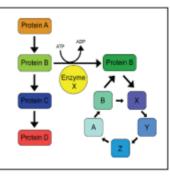
Database Production

Metabolite identification & quantification (NMR/GC-MS/LC-MS)

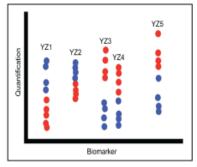
Computational Analysis



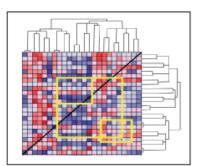
Identify & Compare Classes



Map to Metabolic pathways



Quantify putative biomarkers



Identify metabolite interactions

Conclusion and Future Directions

 The metabolomics enables the identification of the metabolome while revealing tremendous information about cellular function and the metabolic pathways of a cell. It is increasingly used in investigations of more subtle effects, such as the indication of pharmaceutical efficacy, and the probing of life-style changes, nutrition and the complex interconnection of metabolic processes.

Challenges in Metabolomics:

- (1) metabolites have a wide range of molecular weights and large variations in concentration,
- (2) the metabolome is much *more dynamic* than proteome and genome, which makes the metabolome more *time sensitive*,
- (3) metabolites can be either polar or nonpolar, as well as organic or inorganic molecules. This makes the *chemical separation* a key step in metabolomics,
- (4) metabolites have *chemical structures*, which makes the *identification* using MS an extreme challenge.

Future work we will explore:

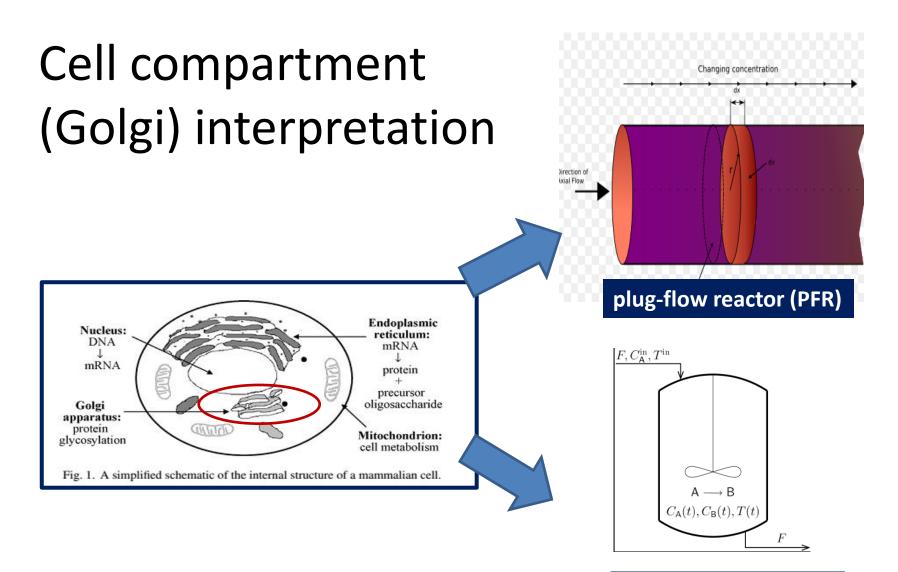
- (1) development of metabolite identification and Standardization of culture conditions,
- (2) standardization of sample preparation protocols will ensure reproducibility and reliability,
- (3) development of new sample preparation methods will increase metabolite coverage in future manufacturing cell metabolomics studies.



Biochemical Reaction Network Modelling for Glycosylation

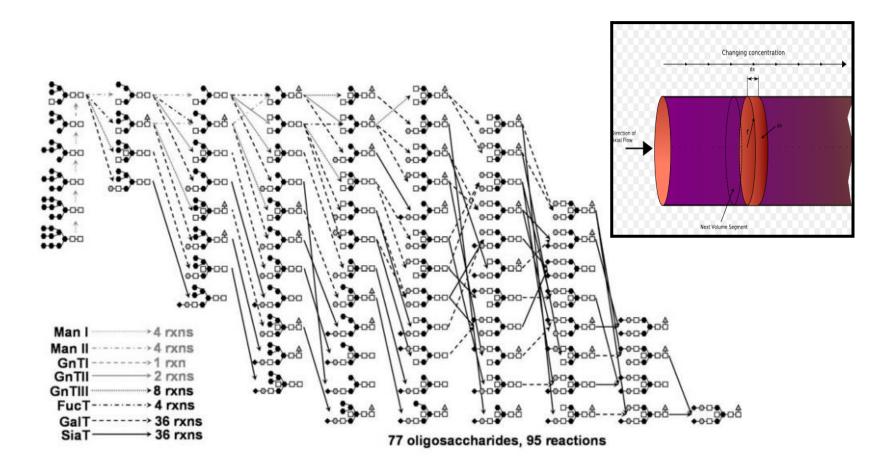
Sha Sha

Biomedical engineering and biotechnology (BMEBT), UMass Lowell



countinous wellmixed reactor (CSTR)

Chemical reaction for glycosylation



Val et al. *Biotechnol. Prog.*, 2011

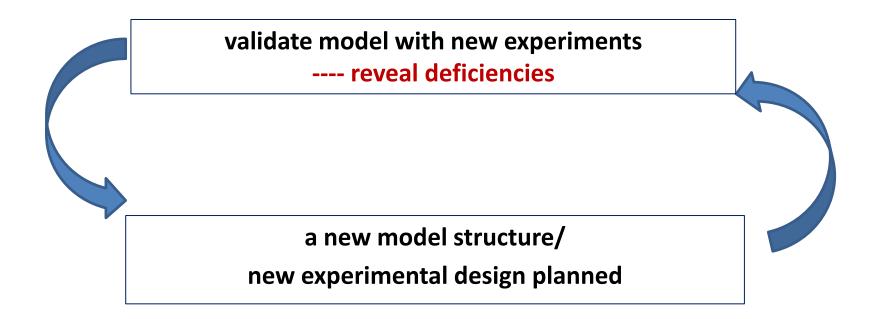
Steps to the first working model

• 1. Define the **frame and structure** of the model based on *a priori* knowledge

• 2. Parameter estimation from available data

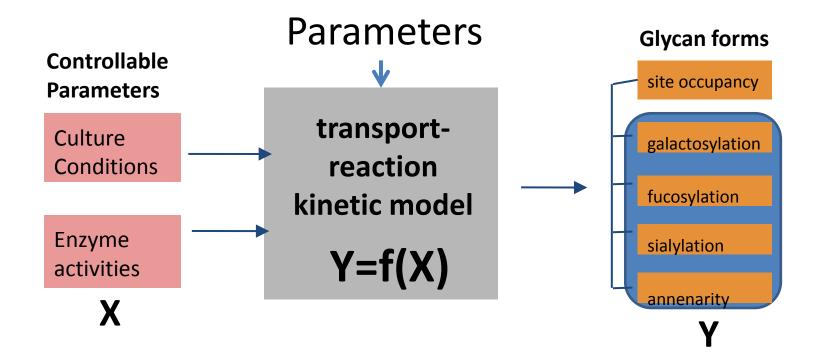
- Golgi size
- Protein flow rate
- Donor concentration (carbon feeding, metabolites)
- Enzyme concentration
- Enzyme kinetic constants
- Transport protein concentration
- Glycan profile
- ..

Validation and Optimization of model



models and experiments are disigned in tandem ensuring that sets of modelled and measured variables can be matched to each other

Conclusion



- Simulate, predict and optimise procedures, experiments and therapies
- Disprove hypotheses and to define improved hypotheses (based on comparison of model-predicted and experimentally measured variables)

Protein A Column: Modeling, Simulation & Multi-Variate Analysis

Ketki Behere

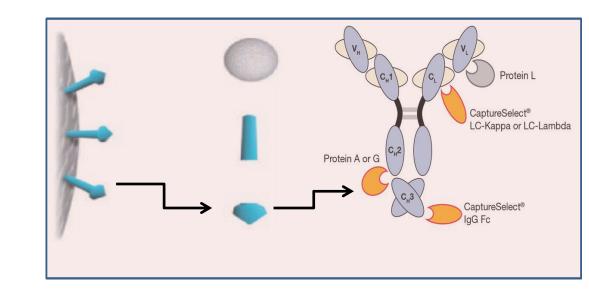
PhD in Chemical Engineering Umass Lowell, MA

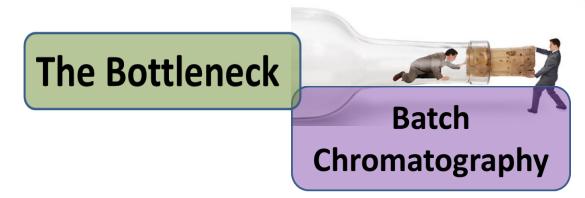
Advisor: Prof Seongkyu Yoon

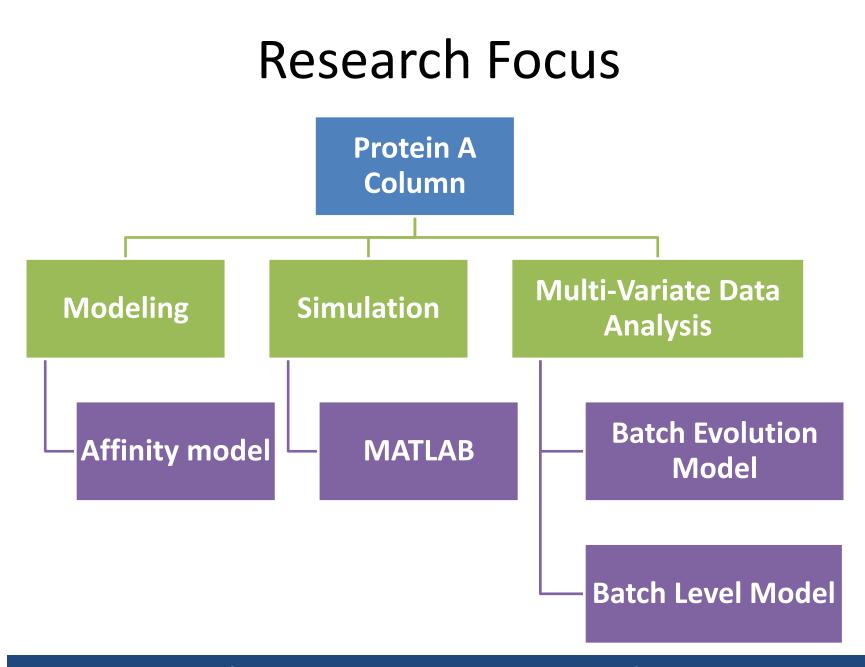


Why Protein A column?

- Protein A: Affinity column used as a capture step
- Affinity binding specific to Monoclonal Antibody provides > 95% purity

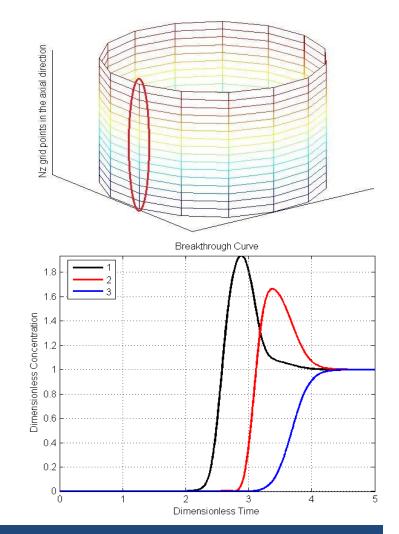






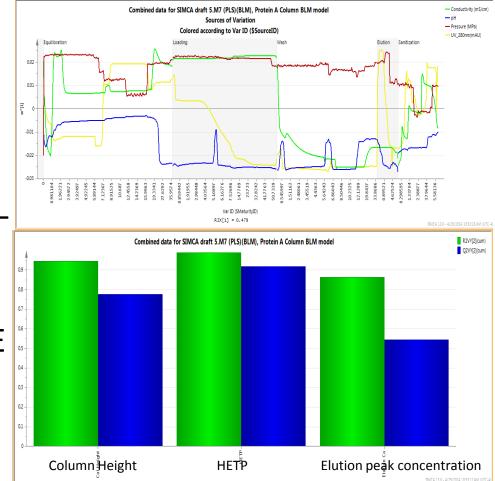
Modeling & Simulation

- General Rate Model customized for Affinity column: Protein A
- MATLAB Simulation for Single Column
- Project Single-column simulation to Multi-Column



Multi-Variate Data Analysis

- Explain the variability in dataset by building a regression model
- Predictability to the user
- Determine the resin shelflife & online qualification of column
- Used to complement DOE studies in process development



Conclusion

- Mechanistic model to better explain, simulate, control and analyze the chromatography process.
- The purification process will become robust improving the process economics and increase the productivity.
- A Continuous Downstream Process (CDP) with Model & Simulation will demonstrate the Next Generation Bioprocess.

