

### The Feedstock-Conversion Interface Consortium – **Understanding and Mitigating the Impacts of Feedstock Variability in Bioconversion Processes**

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October 27<sup>th</sup>, 2022 **ABLCNEXT2022** 



**U.S. DEPARTMENT OF ENERGY** 



## **1-slide guide to the FCIC**

The Feedstock-Conversion Interface Consortium is led by DOE as a collaborative effort among researchers from 9 National Labs

### **Key Ideas**

- Biomass feedstock properties are variable and different from other commodities
- **Empirical** approaches to address these issues have been unsuccessful

The FCIC uses first-principlesbased science to de-risk biorefinery scale-up and deployment by understanding and mitigating the impacts of feedstock variability on bioenergy conversion processes.

http://energy.gov/fcic





















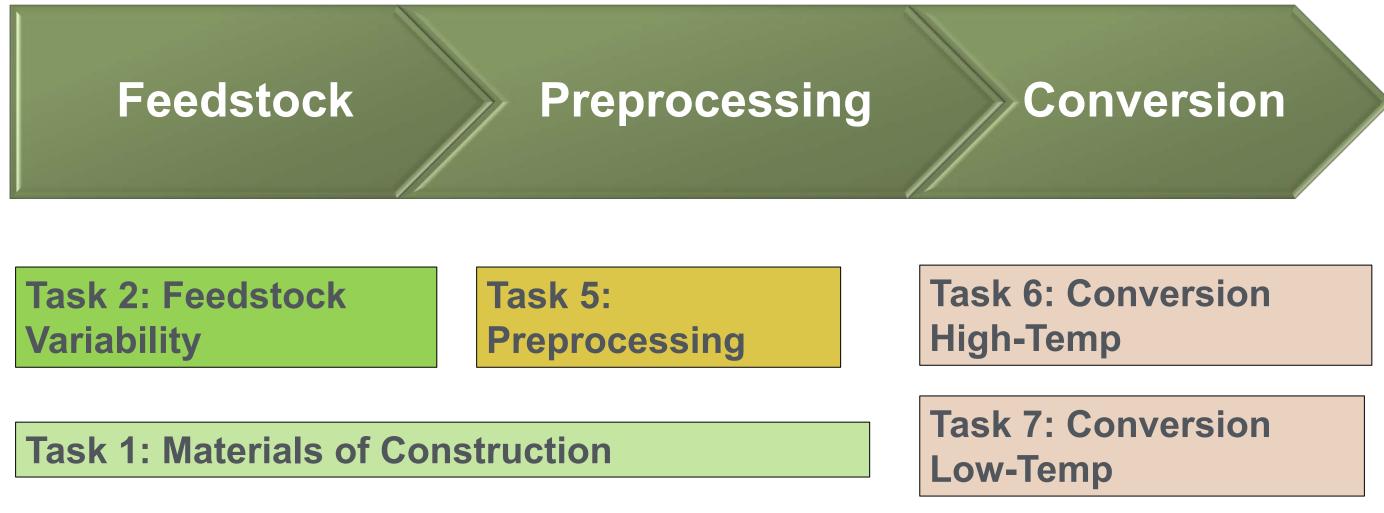








## FCIC Task Organization



**Task 3: Materials Handling** 



Task X: Project Management

**Task 4: Data Integration** 

Task 8: TEA/LCA







Task X: Project Management: Provide scientific leadership and organizational project management

Task 1: Materials of Construction: Specify materials that do not corrode, wear, or break at unacceptable rates

Task 2: Feedstock Variability: Quantify & understand the sources of biomass resource and feedstock variability

Task 3: Materials Handling: Develop tools that enable continuous, steady, trouble free feed into reactors

Task 4: Data Integration: Ensure the data generated in the FCIC are curated and stored – FAIR guidelines

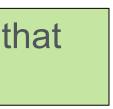
Task 5: Preprocessing: Enable well-defined and homogeneous feedstock from variable biomass resources

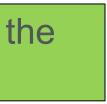
Task 6 & 7: Conversion (High- & Low-Temp Pathways): Produce homogeneous intermediates to convert into marketready products



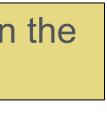
Task 8:Crosscutting Analyses TEA/LCA: Valuation of intermediate streams & quantify variability impact

















### **Quantifying Effects of Variability to Assess Risk**

- An example using low-temperature conversion
  - -Corn stover Feedstock
  - –Deconstruction with Deacetylation & Mechanical Refining and Enzymatic Hydrolysis (DMR/EH)
  - -Upgrading Sugars to mixed organic acids, lignin monomers to muconate
- How does corn stover variability present a risk to biorefineries?





#### Organisms

#### Facilities

#### Products



Rhodosporidium toluloides



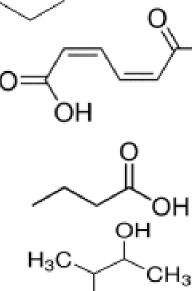




nobilis

Pseudomonas





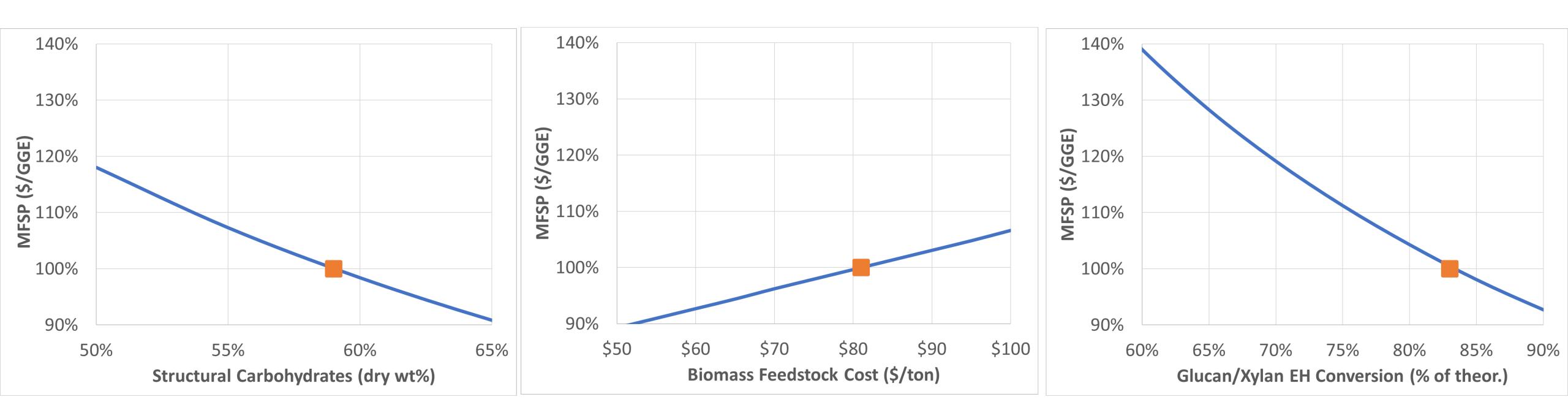






### Feedstock Variability is an Economic Risk

#### **COMPOSITION**





COST

**CONVERTIBILITY** 

### **Risk ~ slope and shape of the lines**

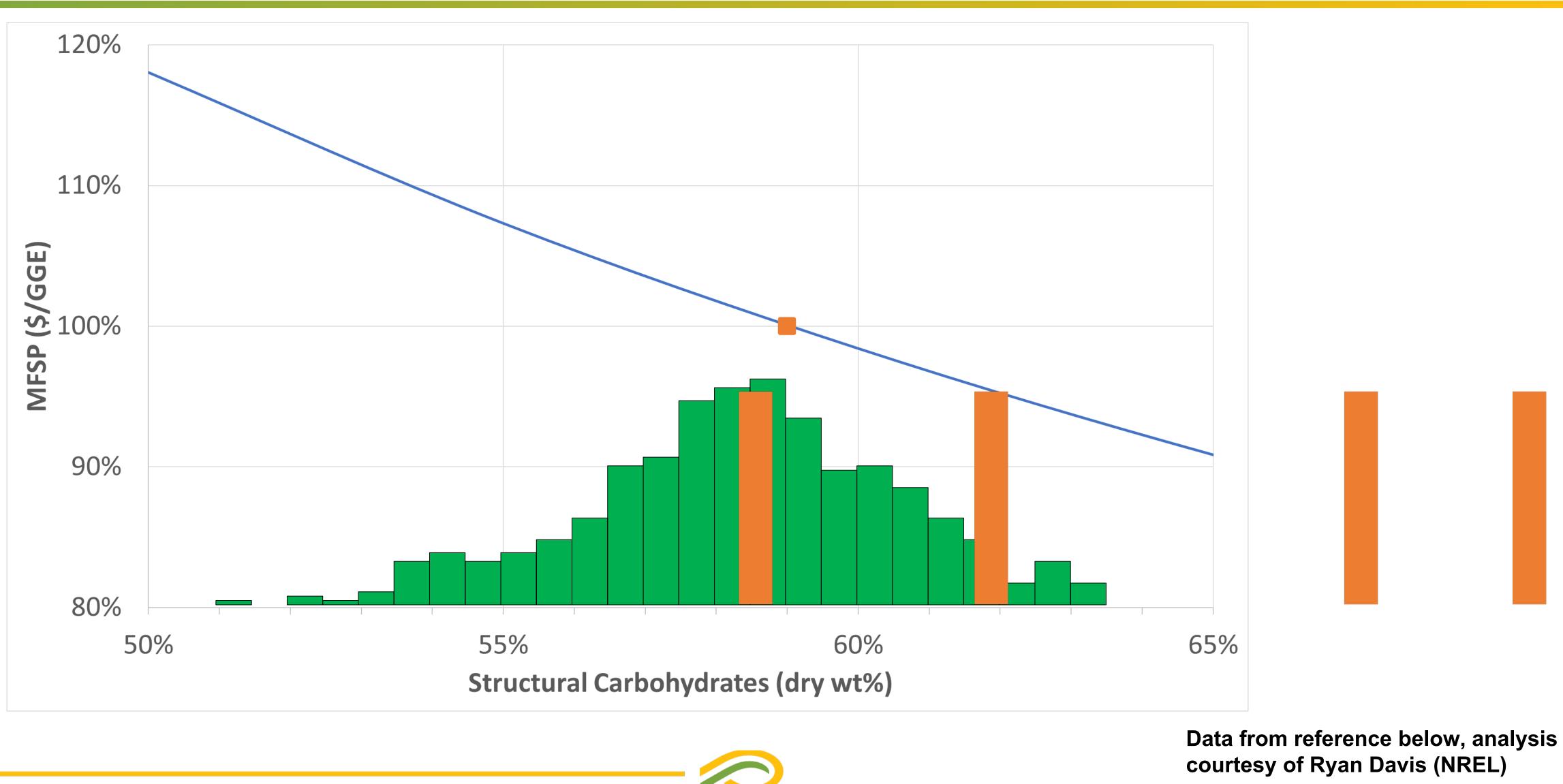


Analysis courtesy of Ryan Davis (NREL)



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### **Composition Variability is a Risk**



Effect of corn stover compositional variability on minimum ethanol selling price (MESP), Bioresource Technology 140:426 (2013), Tao et al, https://doi.org/10.1016/j.biortech.2013.04.083



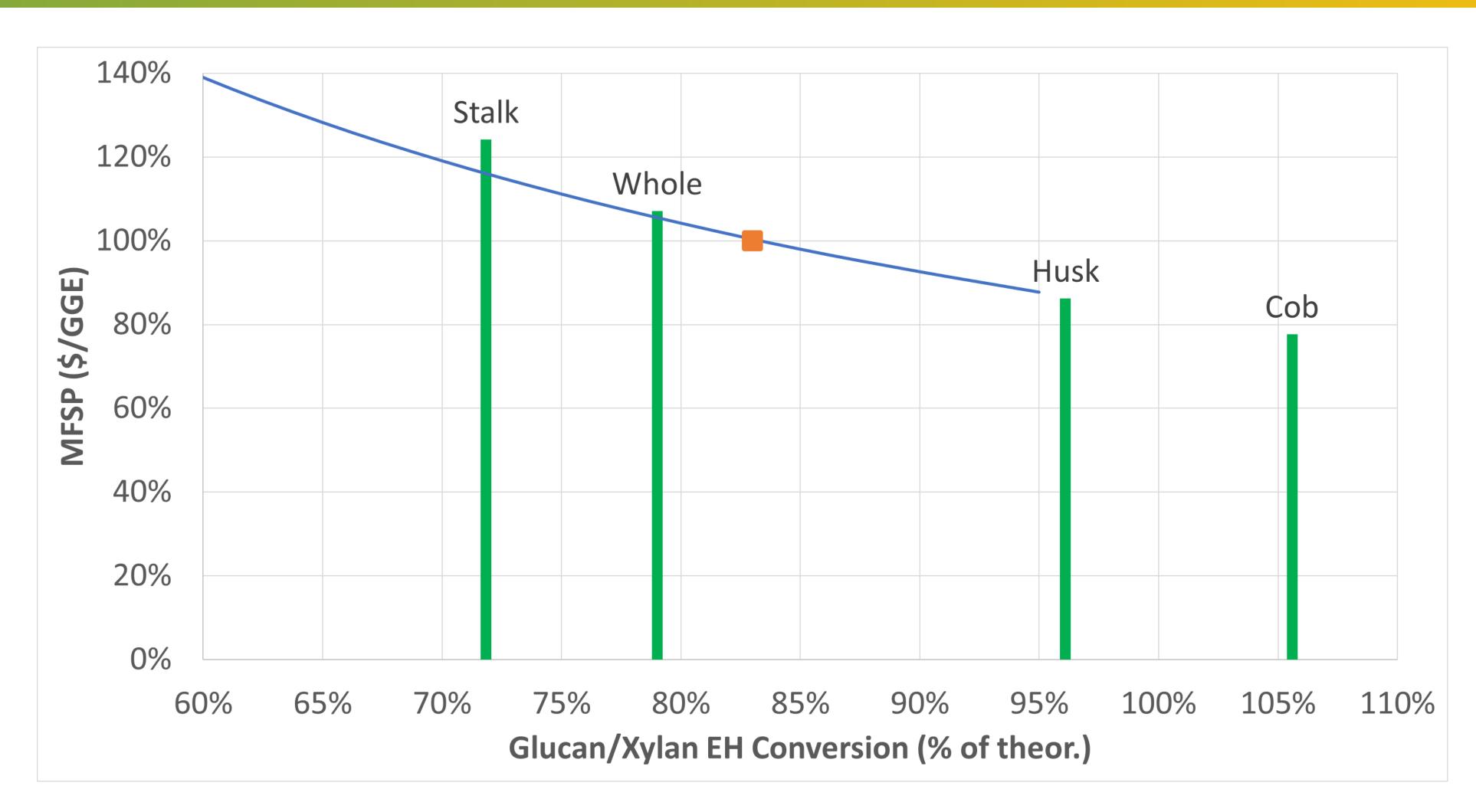








### **Conversion Variability is a Risk**



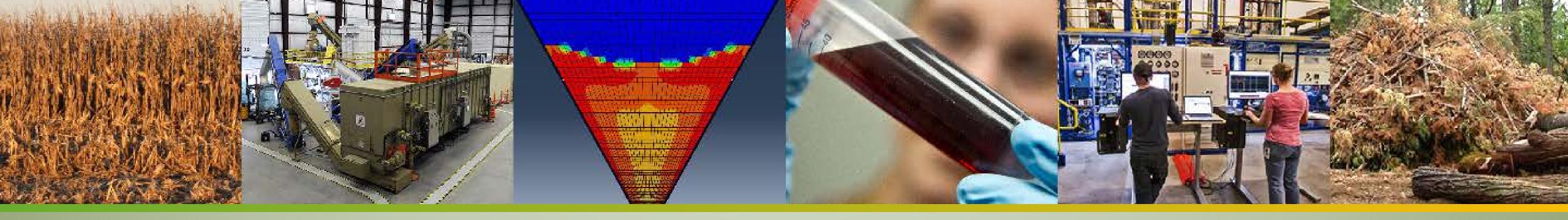
Data courtesy of Xiaowen Chen (NREL), analysis courtesy of Ryan Davis (NREL)











## **Quality by Design**





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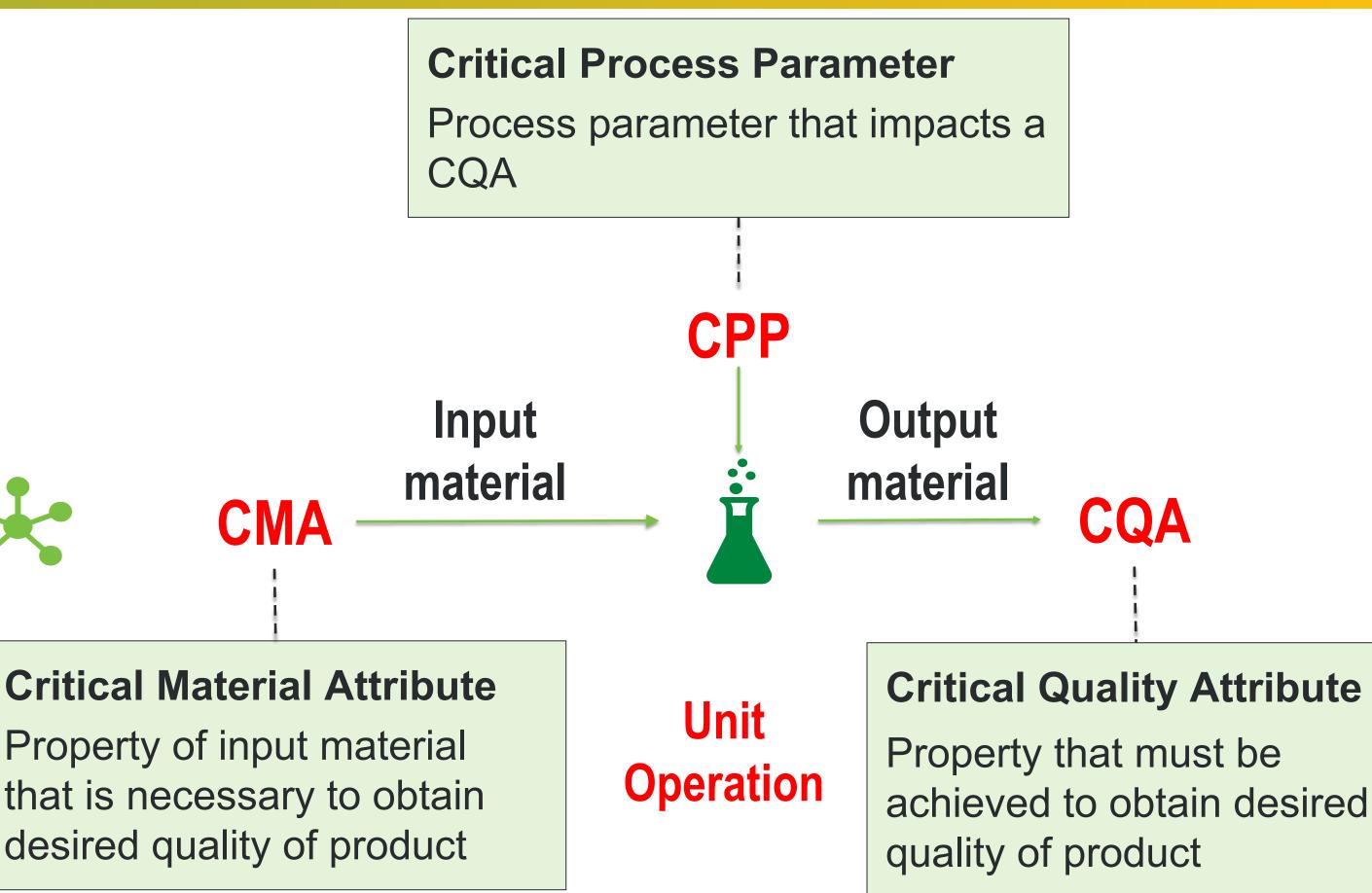


## Quality by Design (QbD) to Assess Risk

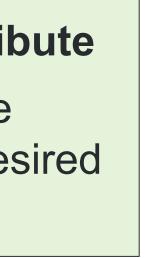
- Key operating concept and organizing principle
- Widely used in pharmaceutical manufacturing FDA-endorsed
- Chemical processes are collections of <u>specific</u> unit operations
- Unit operations are discrete but <sup>+</sup> connected
- Need fundamental understanding of
  - Unit operation
  - Input & Output streams













## **QbD** is about Feedstock Attributes....

# Moving from feedstock NAMES to feedstock ATTRIBUTES









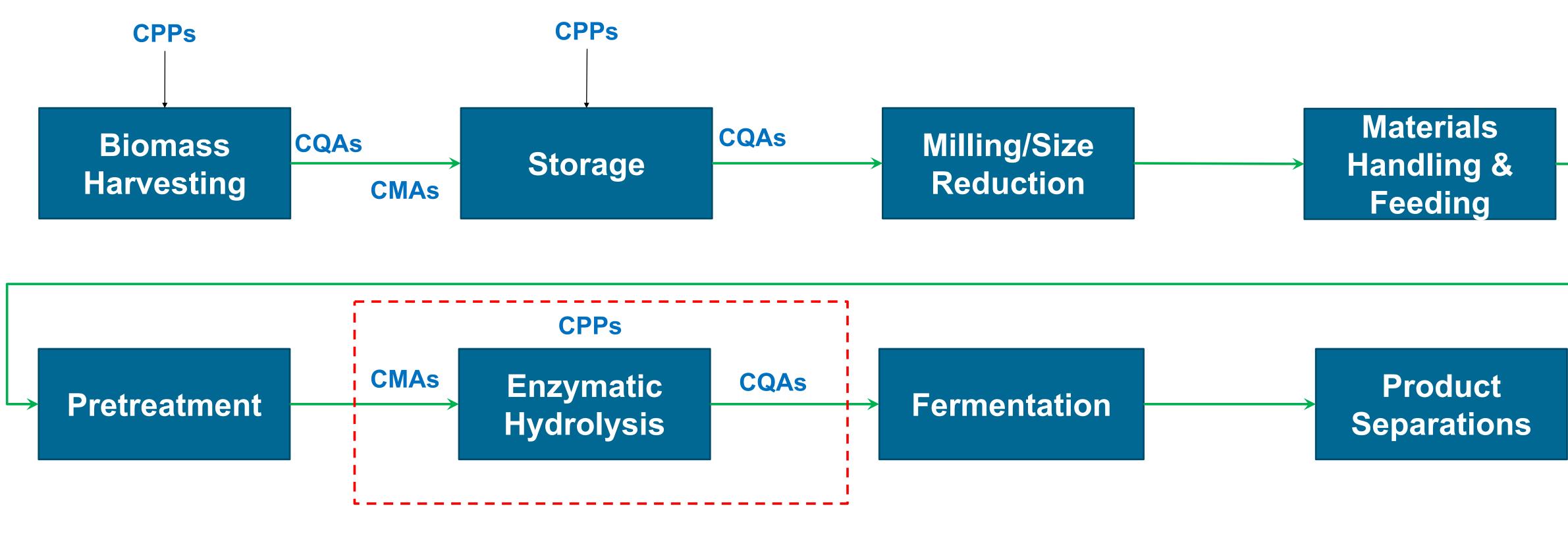




hayandforage.com



### **QbD for the Biomass Value Chain**



CMAs:	<b>CPPs:</b>
<ul> <li>Structural Carbohydrate Content</li> </ul>	<ul> <li>Temper</li> </ul>
Digestibility	<ul> <li>Enzyme</li> </ul>
<ul> <li>Inhibitors (e.g., pretreatment byproducts)</li> </ul>	<ul> <li>Resider</li> </ul>





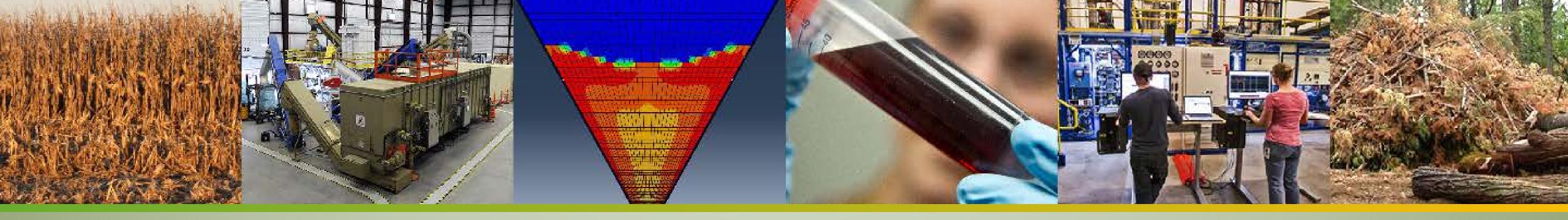
erature ne Loading ence Time

- CQAs:  $\bullet$
- Soluble Sugar Content
- Residual substrates
- Inhibitors





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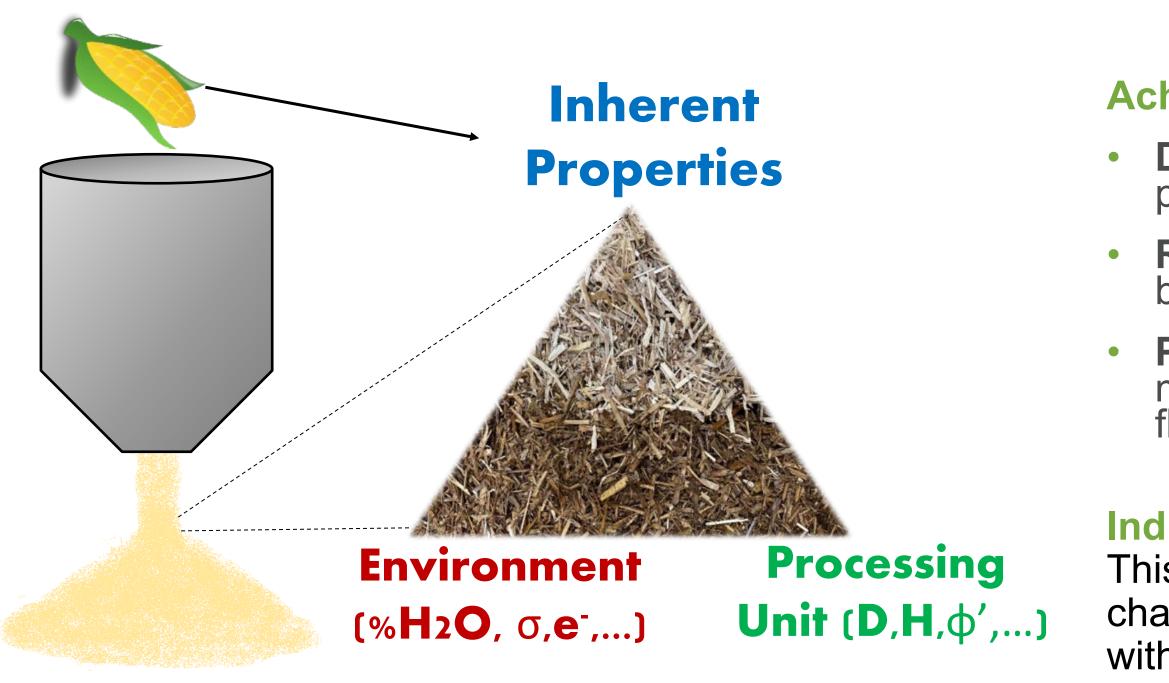
### **Representative Accomplishments**





### **Review Article: Flow Behavior Characterization of Biomass Feedstocks**

It's imperative to test each powder because the flow behavior of a powder is determined by its inherent properties, the environment, and the processing unit it's in.



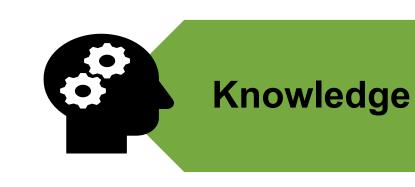














#### **Current Knowledge Gap**

- Existing powder processing/handling equipment are mostly designed for coal and pharmaceutical ingredients, rather than for biomass
- Biomass encounters flow stoppages in these repurposed equipment, hampering process economics
- Lack of knowledge on the unique flow behavior of biomass feedstocks

#### **Achievement**

- **Discussed** how powder flow behavior depends on the inherent properties, the environment, and the processing equipment
- **Reviewed** literature on the characterization of biomass powder flow behavior using shear testers and powder rheometers
- **Proposed** complementing powder rheometry with surface energy measurements, tribometry and DEM modeling to better understand the flow behavior of biomass powders

#### **Industry Impact**

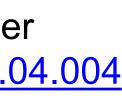
This review article educates audiences working on powder flow characterization in both industrial and academic settings, and provides them with insights on future research directions



Flow Behavior Characterization of Biomass Feedstocks. Powder Technology (2021) 387, 156-180, DOI: <u>10.1016/j.powtec.2021.04.004</u>

#### Task 3 – Material Handling

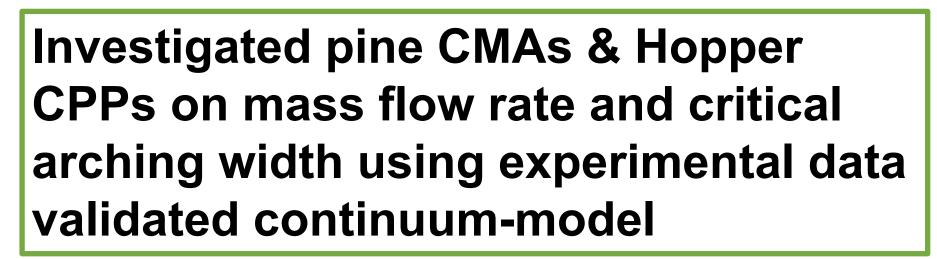


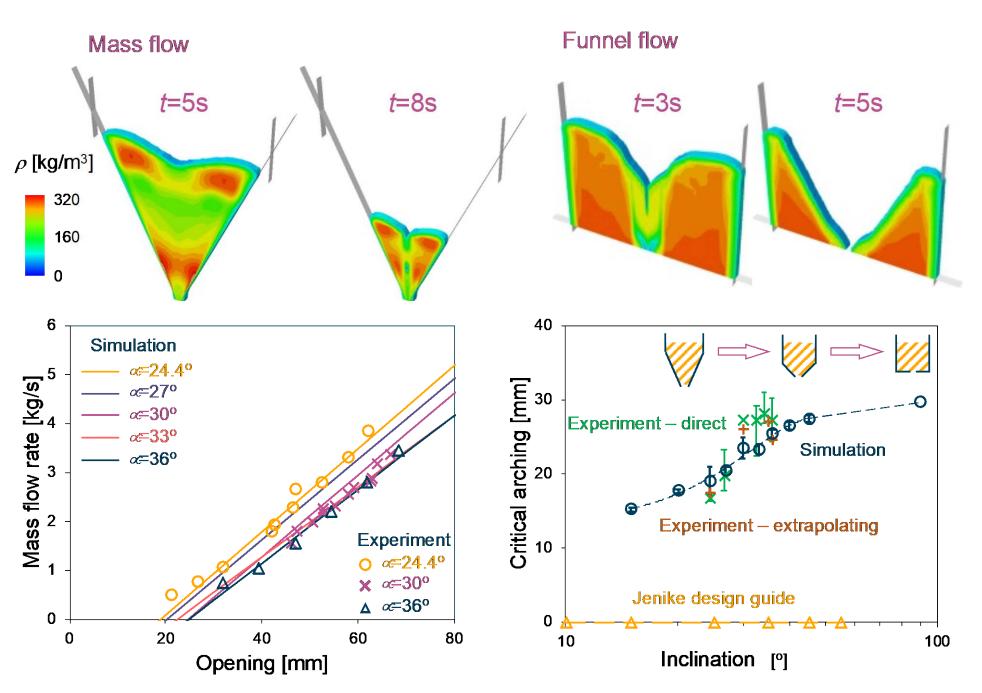






### **Continuum Modeling of Hopper Flow**





Flow and Arching of Biomass Particles in Wedge-Shaped Hoppers, ACS Sustainable Chemistry & Engineering (2021) 9:45, 15303–15314, DOI:10.1021/acssuschemeng.1c05628



### **Current Knowledge Gap**

- The mechanical flow behavior of compressible pine chips is not systematically investigated
- Flowing pine chips through hopper experiences inconsistence manifested as hopper arching, rat-holing, and surging flow

### Achievement

- Conducted **physical experiments** and **numerical simulations** to investigate the influence of pine CMAs and hopper CPPs on hopper flow performance.
- Found hopper outlet width linearly controls the mass flow rate while the hopper inclination angle controls the critical outlet size.
- Found feedstock initial packing determines whether the flow is smooth or surging, and the surcharge-induced compaction creates flow impedance.

### **Industry Impact**

The knowledge can be directly applied to prepare the charging process of wedge-shaped hopper and to operate hopper in terms of inclination angle and outlet opening for handling pine chips.





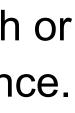


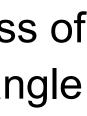








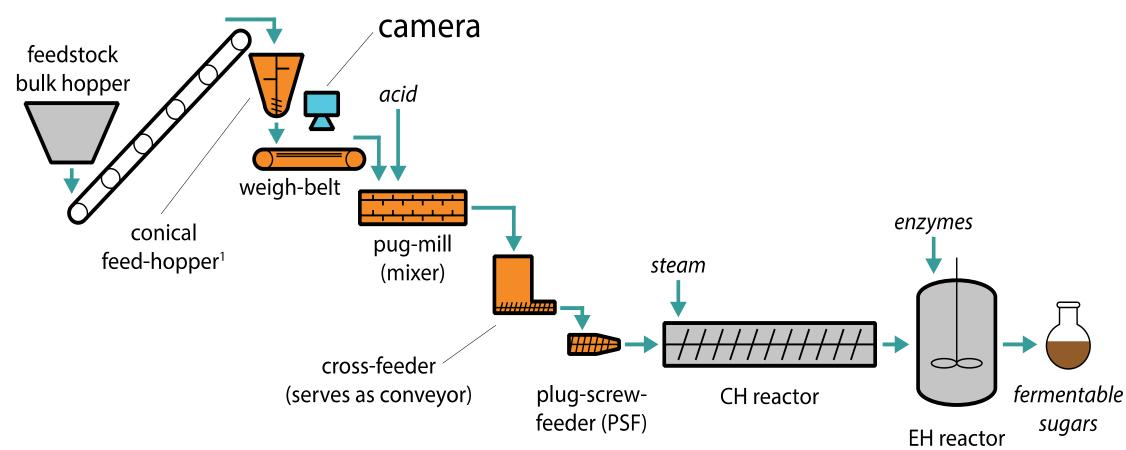






### **Real-Time Feedstock Image Analysis Model**

### Automated machine vision technique detects and quantifies corn stover feedstock particle quality in real-time to enable process control



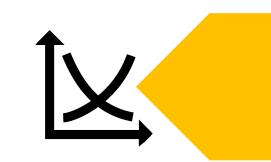
<sup>1</sup>conical feed-hopper cyclically refilled every ~30 min



Real-Time Biomass Feedstock Particle Quality Detection Using Image Analysis and Machine Vision. http://dx.doi.org/10.1007/s13399-020-00904-w











#### **Description**

Utilized a 26,000 image dataset from processing corn stover in a pretreatment reactor captured using **inexpensive digital** cameras.

**Tools** 

Machine Learning Methods - Neural Network (NN) and Pixel Matrix Feature Parameterization (PMFP) used to analyze data

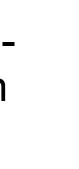
#### Value of new tool

- Neural Network (NN) model can detect anomalies (coarse-particle segregation that can cause feed interruption) even when camera lens obscured by dust.
- PMFP method reveals statistically significant image textural features such as surface roughness, shade variations, and particle angular direction variations that are proxies for particle size distribution variation.
- NN and PMFP approaches are complementary to one another and can describe why feedstock images are classified a certain way.







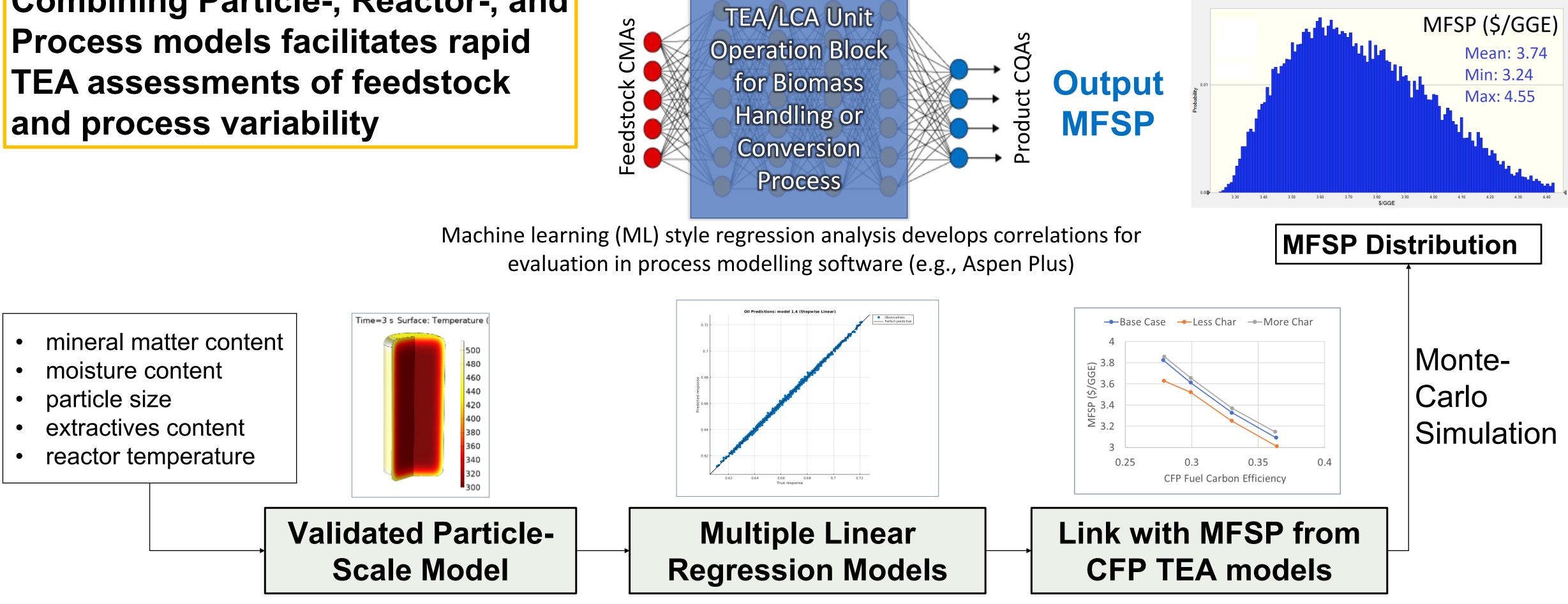






### **Computational Framework For High-Temperature Tools** Modeling – CFD to TEA

**Combining Particle-, Reactor-, and** 

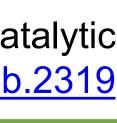


A simplified integrated framework for predicting the economic impacts of feedstock variations in a catalytic fast pyrolysis conversion process <u>https://doi.org/10.1002/bbb.2319</u>

**MFSP**: Minimum Fuel Selling Price; **FP**: Fast Pyrolysis; **CFP**: Catalytic Fast Pyrolysis; **TEA**: Techno-Economic Analysis



Task 6 – High Temperature Conversion



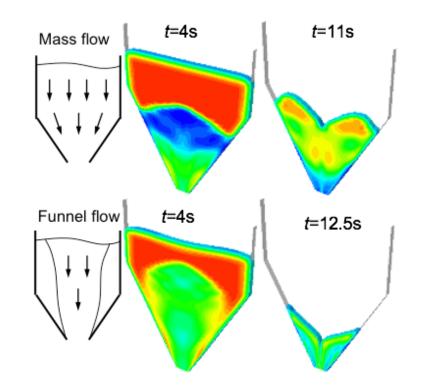


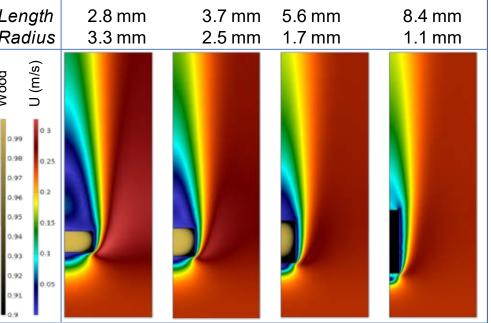
### Key Take-Aways

- Feedstock variability across the Bioenergy Value Chain is a Risk to Biorefineries
- FCIC Researchers are using elements of the Quality-by-Design approach to understand and mitigate the impacts of **feedstock** variability on bioenergy conversion processes.
- Deep subject matter expertise, detailed chemical, physical, and mechanical characterization, and robust and validated modeling is providing knowledge and tools to bioenergy stakeholders

















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This work was authored [in part] by the National Renewable Energy Laboratory, operated by Alliance for Sustainable Energy, LLC, for the U.S. Department of Energy (DOE) under Contract No. DE-AC36-08GO28308. Funding provided by the U.S. Department of Energy Office of Energy Efficiency and Renewable Energy Bioenergy Technologies Office. The views expressed in the article do not necessarily represent the views of the DOE or the U.S. Government. The U.S. Government retains and the publisher, by accepting the article for publication, acknowledges that the U.S. Government retains a nonexclusive, paid-up, irrevocable, worldwide license to publish or reproduce the published form of this work, or allow others to do so, for U.S. Government purposes.

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