

Technological, Climate Change and Sustainability Aspects of Future Transportation Fuels

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California Biomass Collaborative

Ethanol Today:

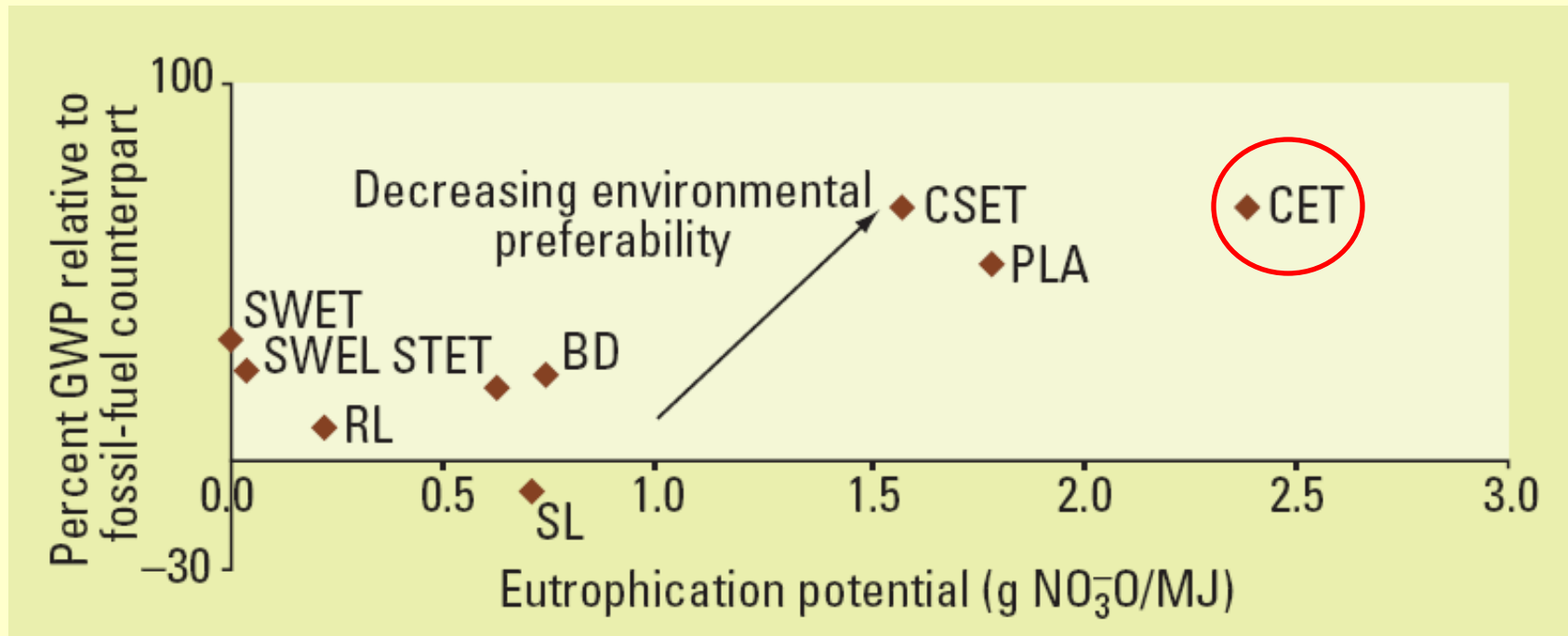


- Currently > 6 million flexible fuel vehicles (US/Can)
 - Run on up to 85% ethanol (by volume)
- Produced from corn grain, *starch*-based technology
 - Co-products key in production economics
 - US: 7.8 billion gal in 2007 (2% energy in transportation)
- Future expansion limited to 15-18 billion gal (corn)
- Policy initiatives to stimulate advanced biofuels
 - U.S. EPA Act 2005; EISA, 2007: 21 billion gal *cellulosic* ethanol/biofuel by 2022, beginning in 2016
 - California: Low Carbon Fuel Standard

Ethanol: Energy and Environment

- Over its life cycle, compared to gasoline, corn ethanol:
 - Significantly reduces petroleum use (~95%), moderately lowers (13%) fossil energy use (Farrell et al. 2006);
 - potentially **increases** overall GHG emissions due to indirect land use change (Searchinger et al., 2008)
 - GHG reduction and co-product credits (Groode and Heywood 2006)
- Only *lignocellulosic ethanol* offers large reductions in fossil energy use/GHG emissions **AND** large production volume (Farrell et al. 2006)
- Lignocellulosic ethanol LCAs:
 - Sheehan et al. (2004); Spatari et al. (2005); GREET (Argonne)
- **Research Gaps:**
 - Many ethanol conversion technologies
 - Variation in technological and “sustainability” performance
 - Uncertainty

Conflicting Sustainability Criteria



Sustainability issues:

Sustainability criteria¹	
Ecological	Socio-economic
Water use	Food and energy security
Water pollution	Land tenure
Organic pollutants	Net Employment
Agro-chemicals	Income distribution
Biodiversity	Wages
Soil erosion	Working conditions
Fertilizer use	Child labor
GMOs	Social responsibility
GHGs/energy input	Competitiveness
Harvesting practices	Culture - Traditional way of

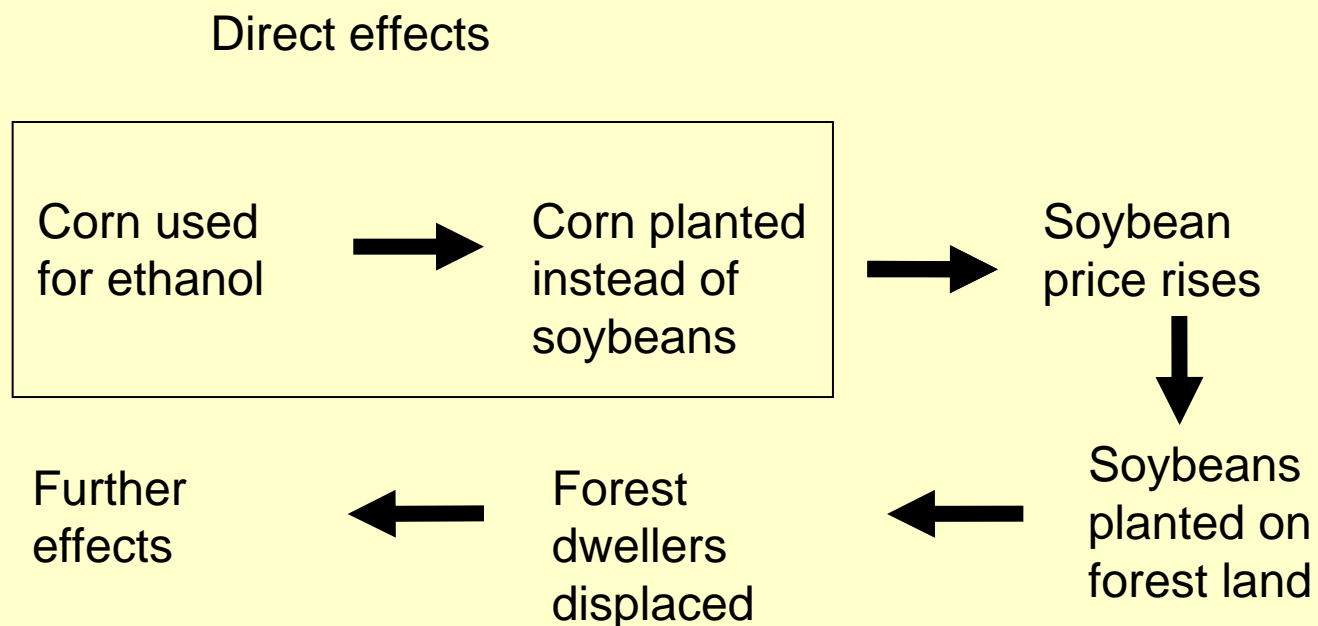
life

¹Direct + Indirect

Scale: Regional, national, global

Market Mediated Effects:

- Indirect land use change (LUC) scenario:



- Use computational general equilibrium model (CGE) to determine indirect LUC, but this is far from actual sustainability measures

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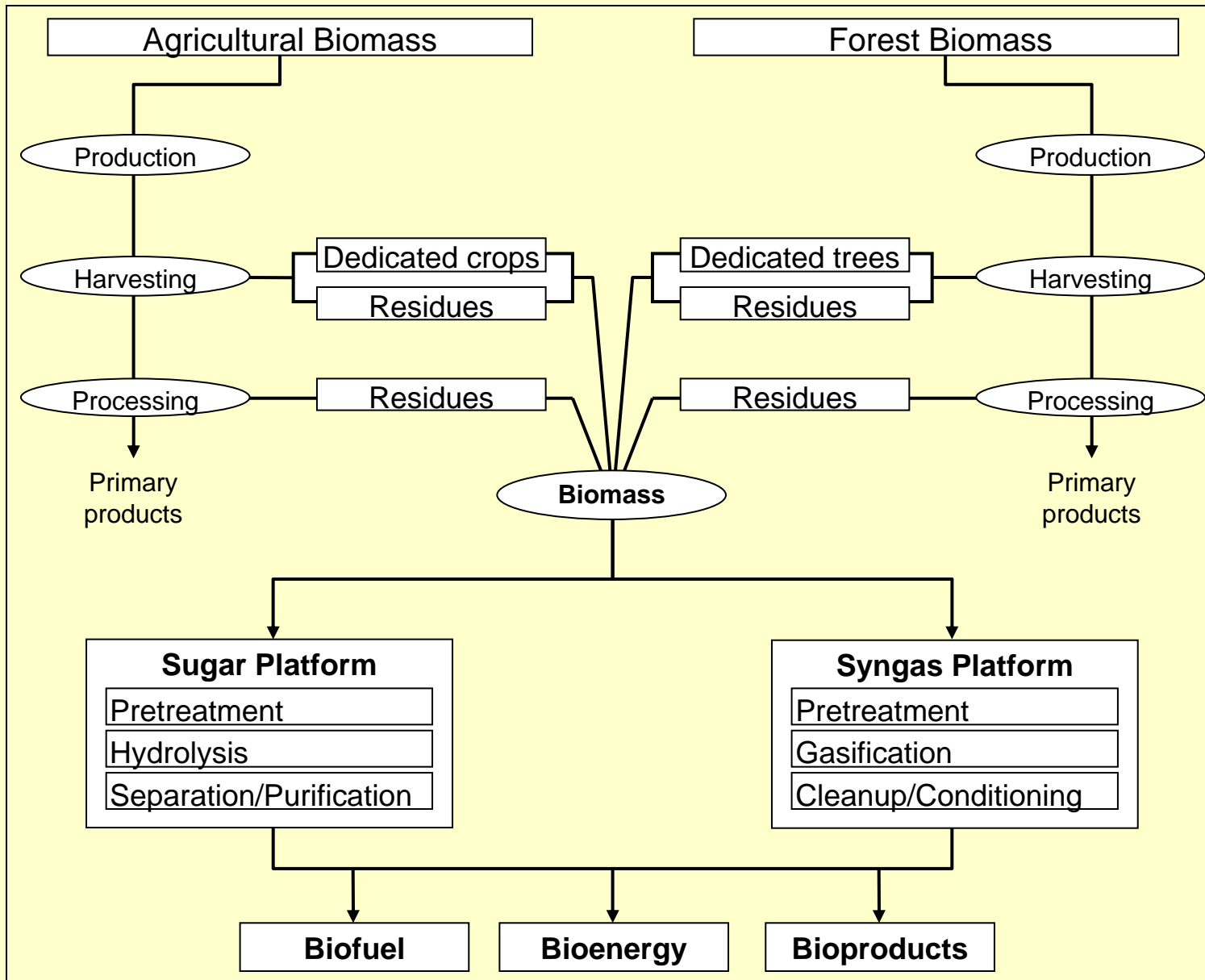
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Biorefinery Production Pathways



Lignocellulosic Ethanol

- Lignocellulosic feedstocks contain
 - Cellulose
 - Hemicellulose
 - Lignin

Energy crops (e.g., switchgrass), agricultural and forest/ mill residues, municipal solid waste

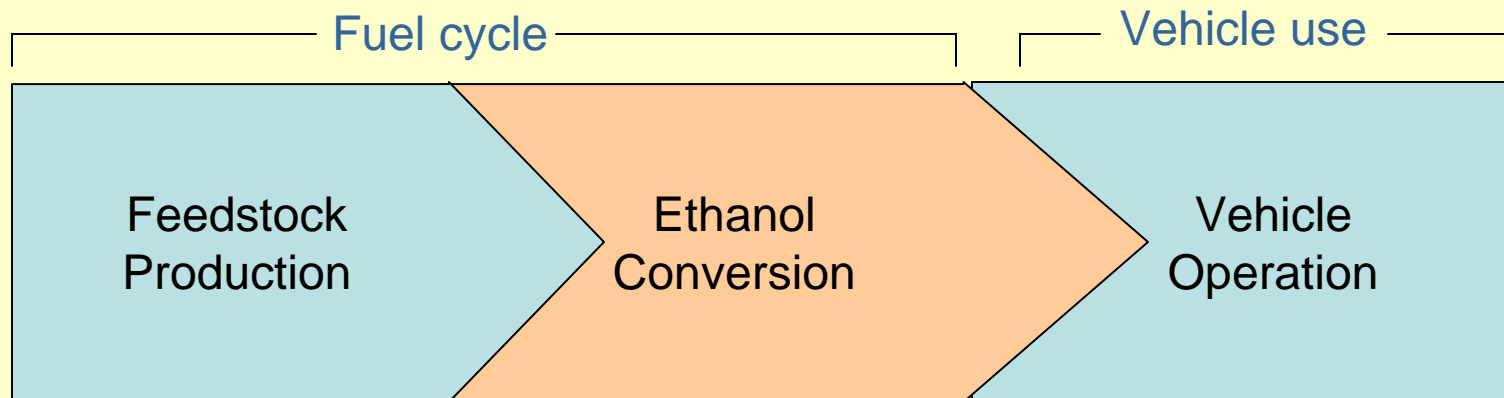
- Advanced biotechnology performs hydrolysis then fermentation on cellulose/hemicellulose fraction
 - Lignin portion of biomass utilized for energy (co-product)
- **Not yet at commercial scale**



Technological Challenges

- Key challenges for R&D:
 - Overcoming the “recalcitrance” of the cellulosic feedstock (Stephanopolous, 2007)
 - Improving enzyme performance
 - Improving enzyme specific activity (FPU/g cellulase)
 - Reducing enzyme costs
 - Reducing pretreatment chemical costs (Himmel et al, 2007)
- Result in improved yields, better cost performance
- But, still much uncertainty in performance

Life Cycle Model



- Fertilizer
- Herbicides
- Harvesting operations

Feedstocks:

- Corn stover (CS)
- Switchgrass (SG)
- Douglas fir (Df)

- Pretreatment chemicals
- Enzymes
- Nutrients

Technologies:

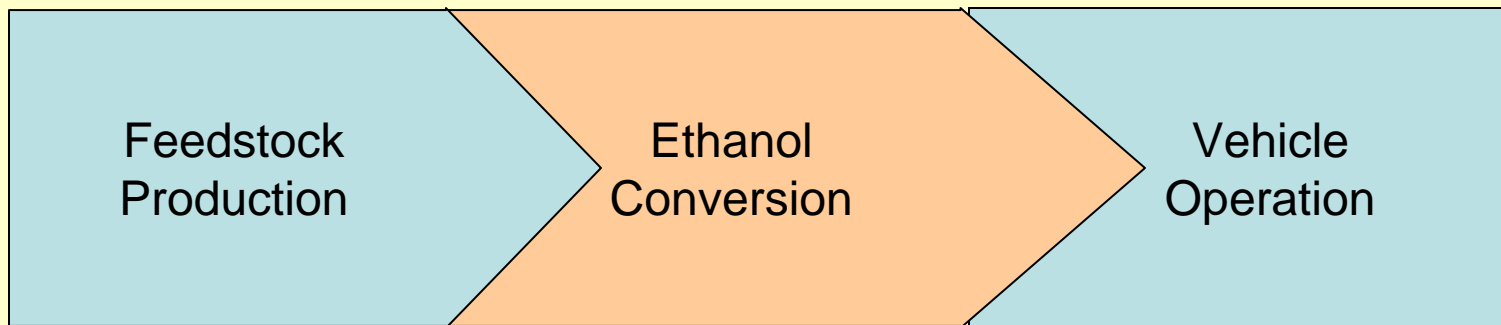
- Dilute acid (NREL)
- Ammonia fibre explosion (AFEX)
- Steam explosion (SE)
- Organosolv (OS)
- Enzymatic Hydrolysis
- Fermentation

- Blending with gasoline
- Vehicle operation

Vehicles:

- Ethanol-fueled vehicle (E85)
- Reformulate gasoline-fueled vehicle (RFG)

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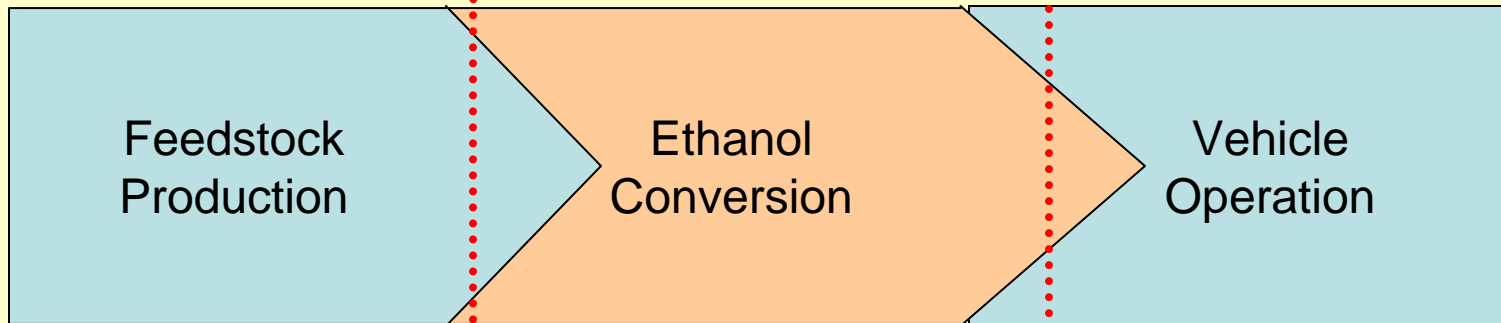
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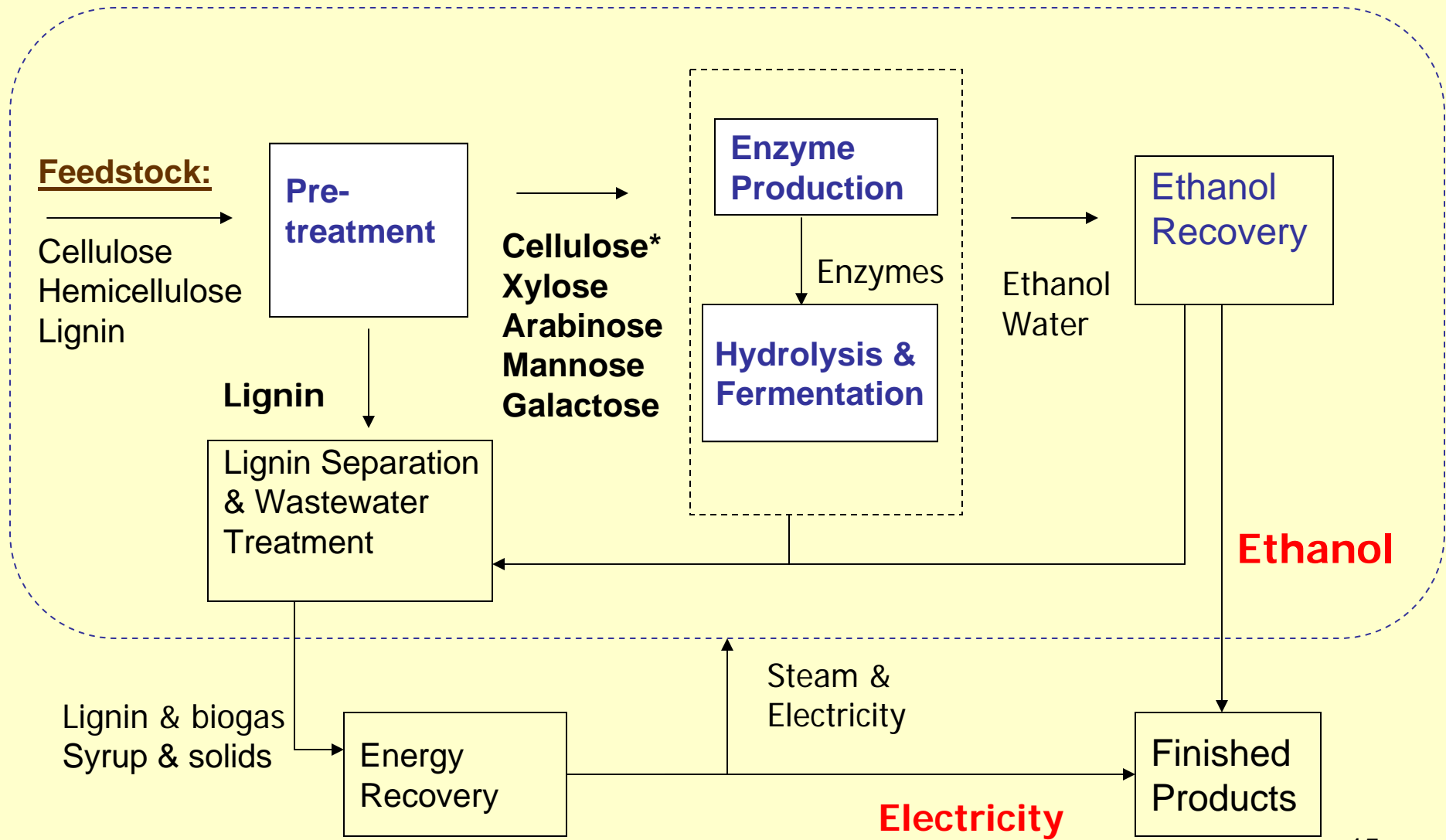
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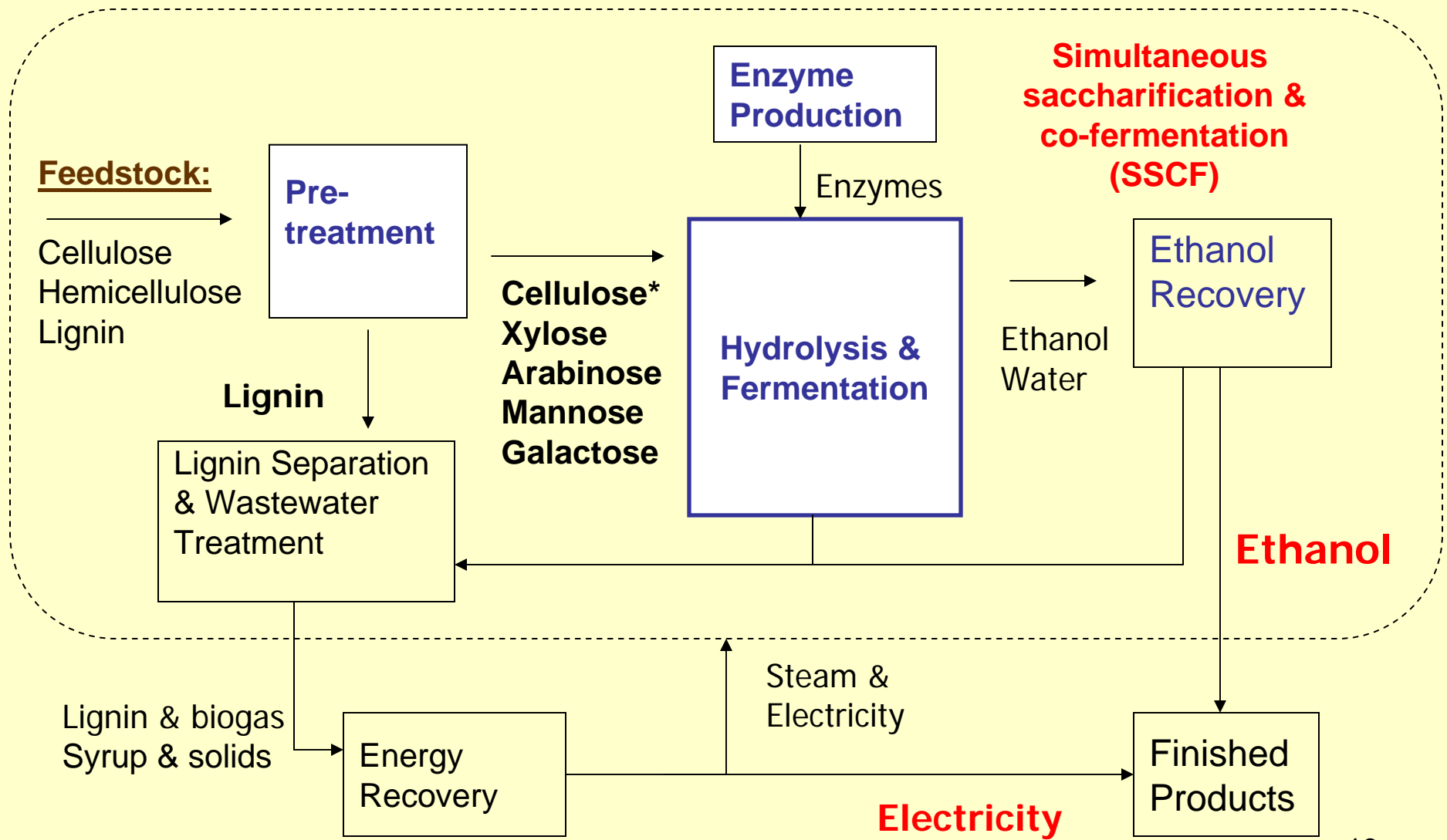
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Ethanol Conversion Model



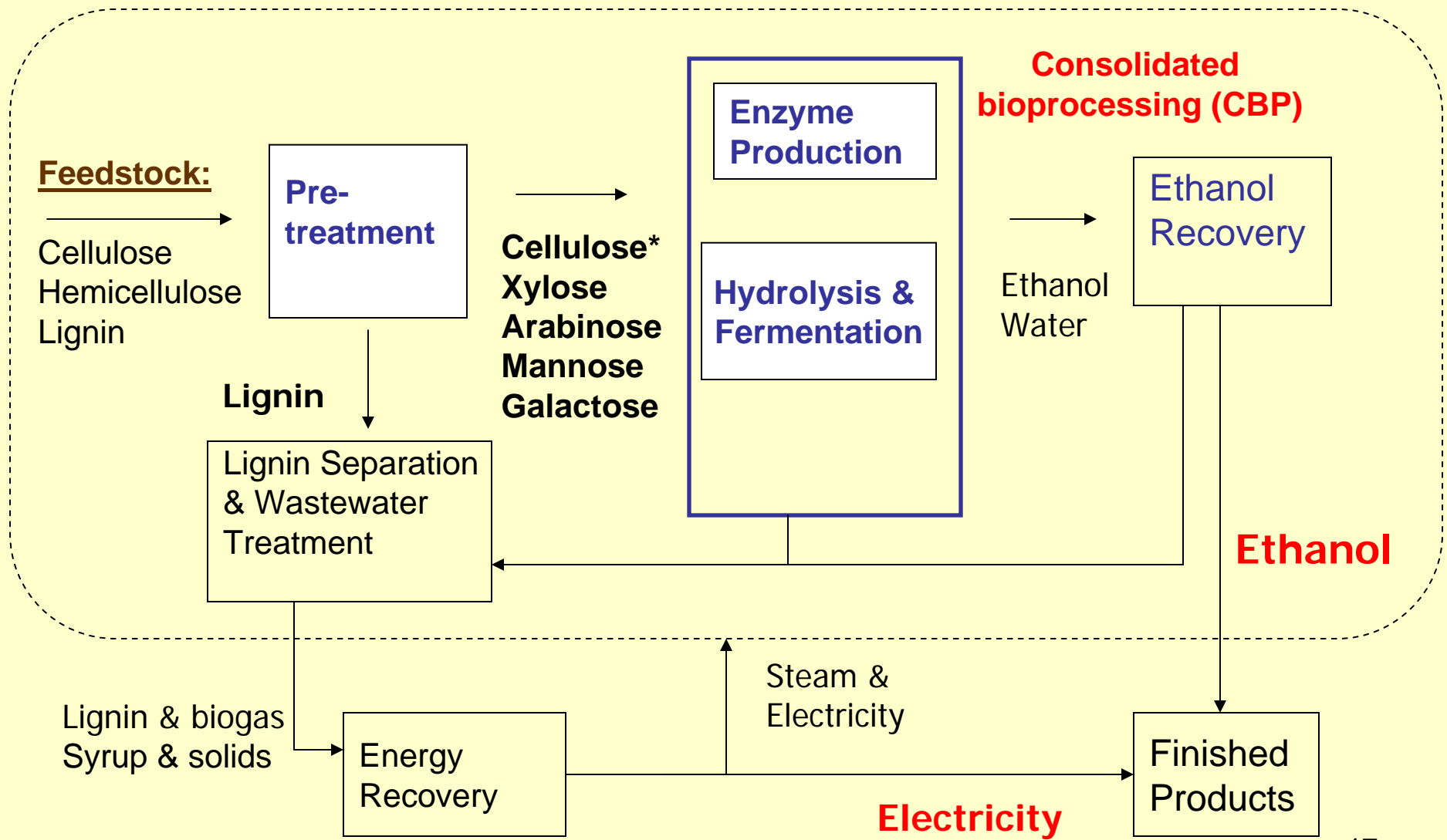
* Pre-treated cellulose

Ethanol Conversion Model: Near-term



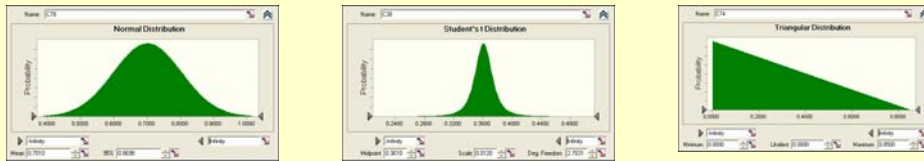
* Pre-treated cellulose

Ethanol Conversion Model: Mid-term



* Pre-treated cellulose

Model Equations & Variables: Performance



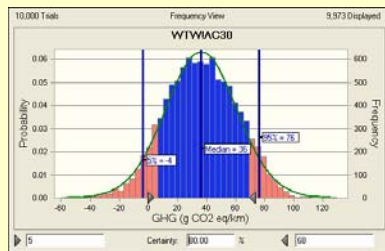
x_1 y_2 $y_3 \dots$

Ethanol (Y_i) = $f(x_1, x_2, x_3; y_1, y_2 \dots)$
 Electricity (E_b) = $g(x_1, x_2, x_3; y_1 \dots)$

Ethanol yield, Y_i (L/metric ton)

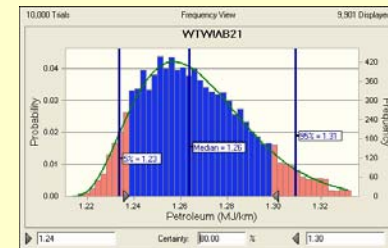
Electricity capacity, E_b (MW)

Sample model results



9 ethanol conversion variables:

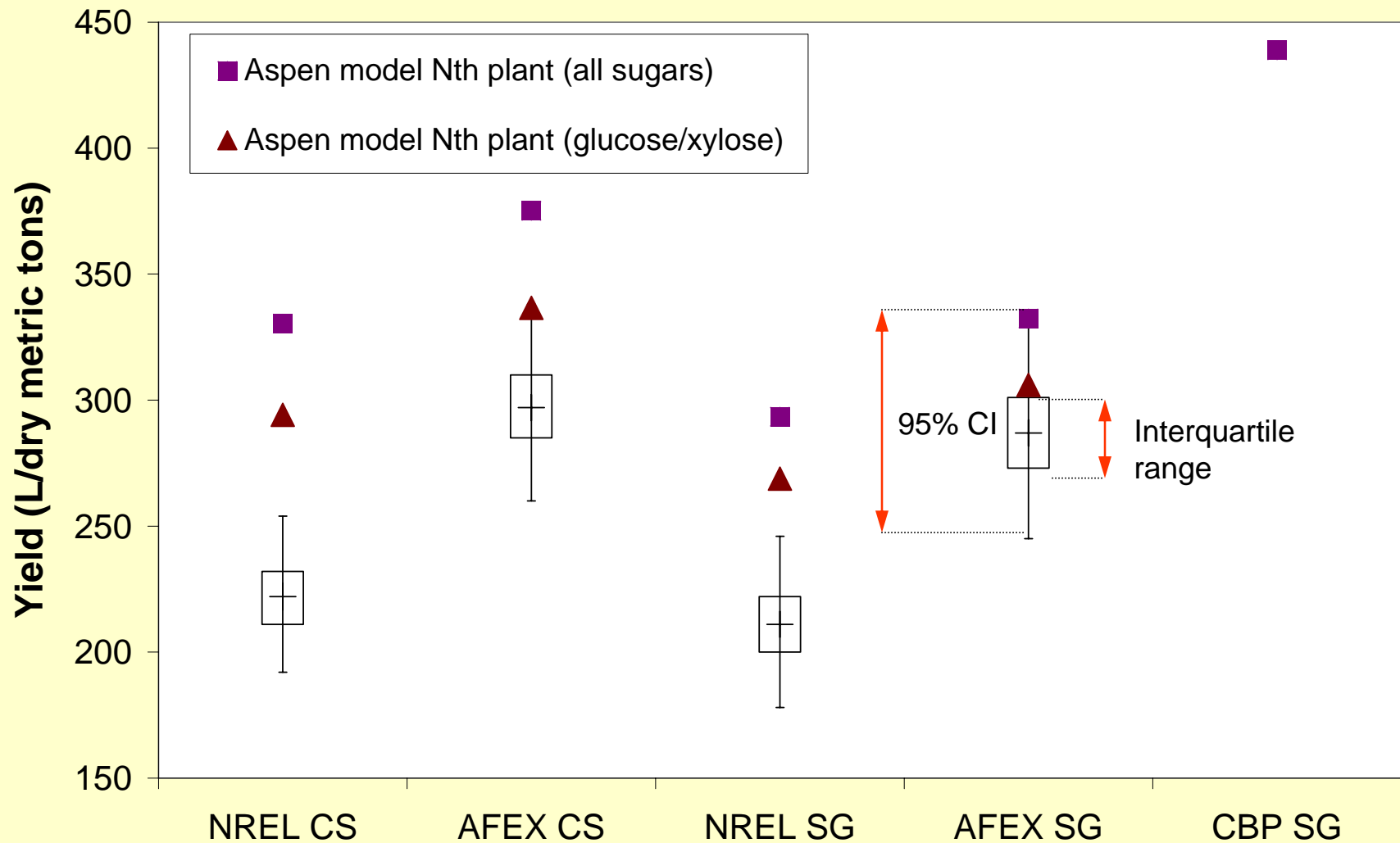
- Feedstock (2)
- Pre-treatment (1)
- Hydrolysis (1)
- Fermentation (5)



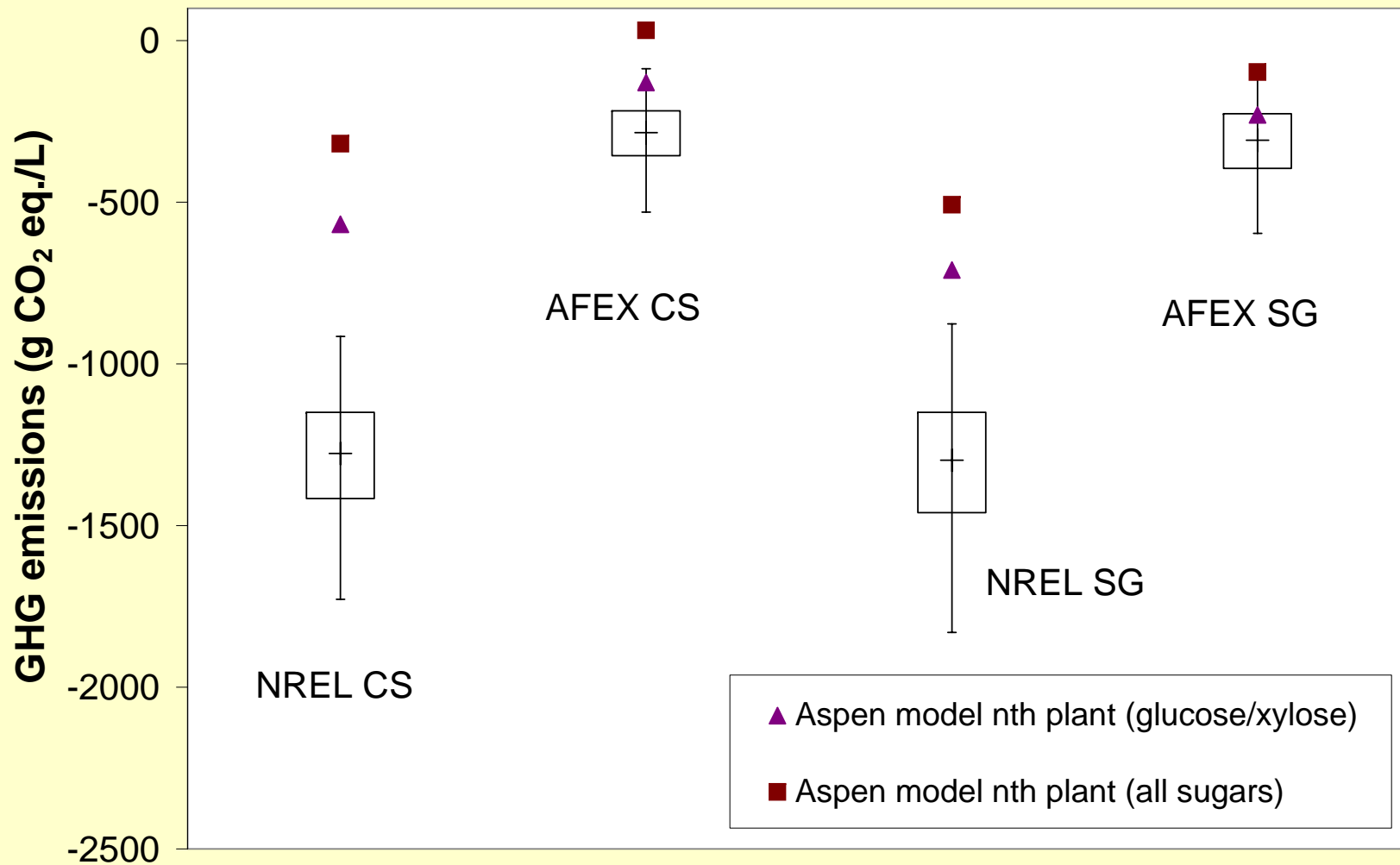
$$Y_i = \sum_i \sum_j \sum_k x_i \times y_j \times \beta_k$$

$$E_b = \sum_{i=1}^n M_i LHV_i - M_F \cdot LHV_E \sum_{i=1}^5 Y_i$$

Performance: Ethanol Yield

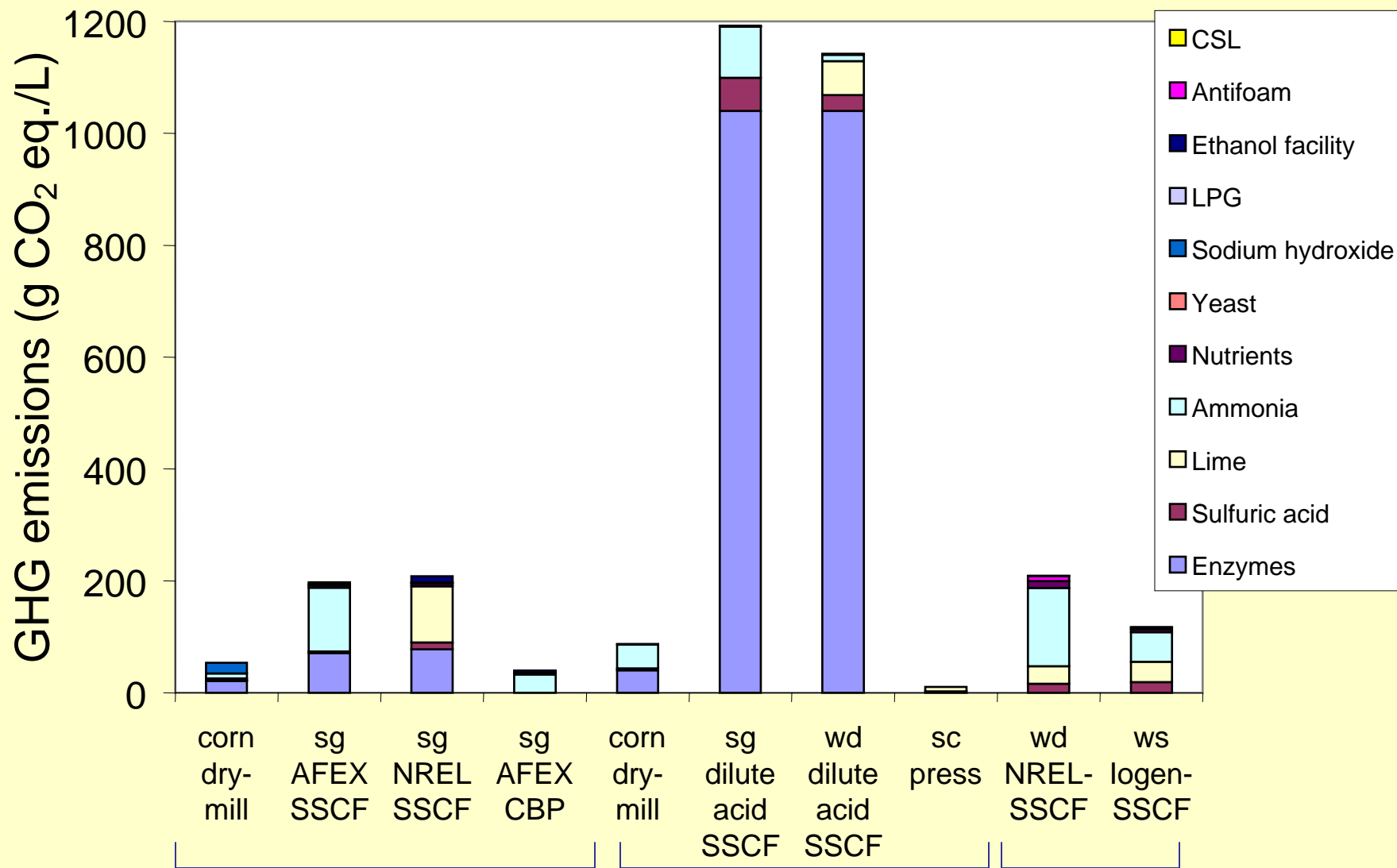


Environment: WTG GHG emissions



Electricity credit included

Ethanol Conversion: Chemicals and Enzymes



Spatari and MacLean

NRCan

EUCAR

Results Depend on Input Variables!

- Change chemical or enzyme loading requirements
 - Vastly increases GHG emissions
- Case illustrated: Cellulases are still specialty products, only a few decades old, high production costs
 - Endoglucanases, exoglucanases, β -glucosidases
 - Technology still evolving
- Need for plausible probability distributions for all significant variables in any LCA model
- Important to update results with new information, with technological change
 - Bayesian updating techniques useful

How to estimate/evaluate S metrics

- Option 1: Qualitative criteria
 - Best practices = “Good”
- Option 2: Estimate a scalar sustainability (S) metric:
 - Compute all sustainability measurements in some “S” unit, social cost-benefit?
 - e.g., define 1 ha biodiversity loss = X “S” units
 - OR
 - Ranking system, ordering importance of S criteria:
 - Biodiversity = 40 units S
 - Cultural diversity loss = 20 units S

How to estimate/evaluate S metrics

- Option 3: Binary system to evaluate feedstocks/technologies:
 - “acceptable” – MSW feedstocks
 - “not acceptable” – feedstocks grown on arable land
- Option 4: Define a vector of mixed “S” criteria
 - Set threshold levels for each S criteria
- Option 5: Combinations of all of the above
- Very difficult to define, measure, and evaluate sustainability of fuels!

Challenge to Calculating Sustainability (S)

- Treatment of uncertainty across S criteria
 - Compare stochastic model results for GWI with the point estimates and/or undefined S criteria
 - Hardly close to estimating uncertainty for S AND it is important for decision making!
 - Also important are sustainability criteria that may conflict with GWI
 - Eutrophication versus GHG reduction
- Maybe a qualitative evaluation approach is best for now
 - Encourage sustainable production practices

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- Bruce Dale, Michigan State University



Life Cycle GHG Emissions

Without electricity credit

