

Sustainable Process Synthesis and Intensification for Industrial Decarbonization

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Experiment Station

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November 09, 2022

Chemical Process Industry (CPI)



Meets the societal demands for chemicals and commodity products

Raw materials

Fossil fuels (**coal**, **natural gas**, and **petroleum**), air, water, salt, limestone, **sulfur** or an equivalent, and some specialized raw materials.



Chemical enterprises



Products

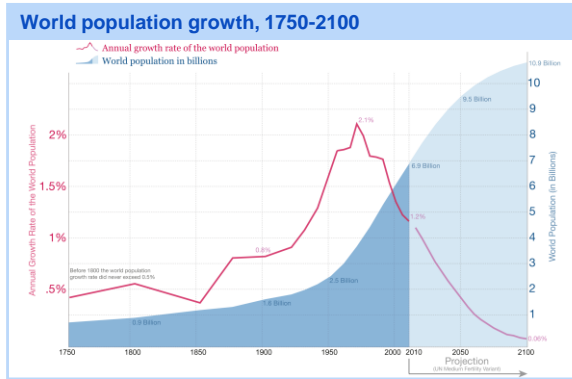


- In 2017, the chemical industry contributed **\$5.7 trillion** to global GDP
- In 2018, **top 50 chemical companies**, excluding PetroChina, had a combined chemical revenue of **\$926.8 billion** (C&EN, July 2019)

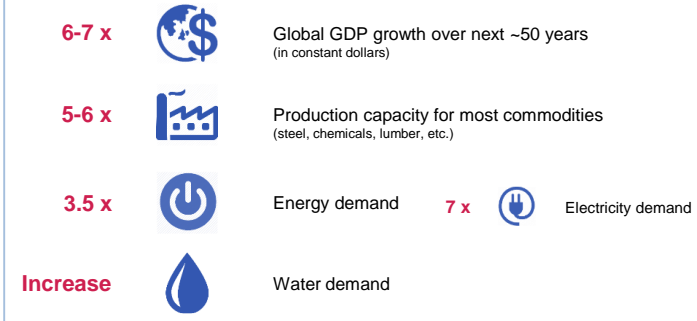
Global Challenges



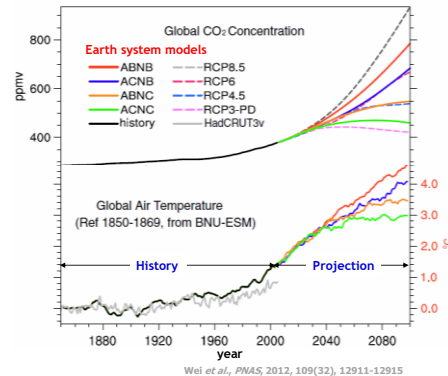
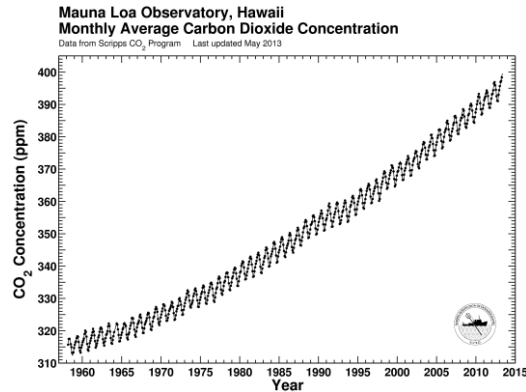
Increasing demand



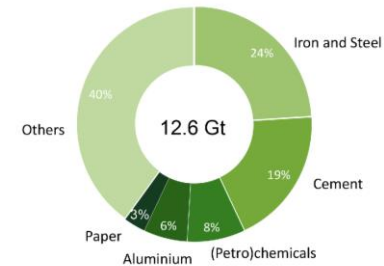
Increase in water, energy & commodities demand



Increasing emission

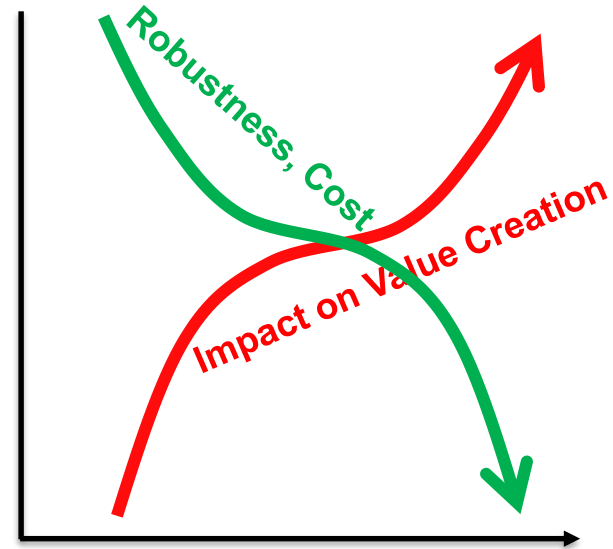
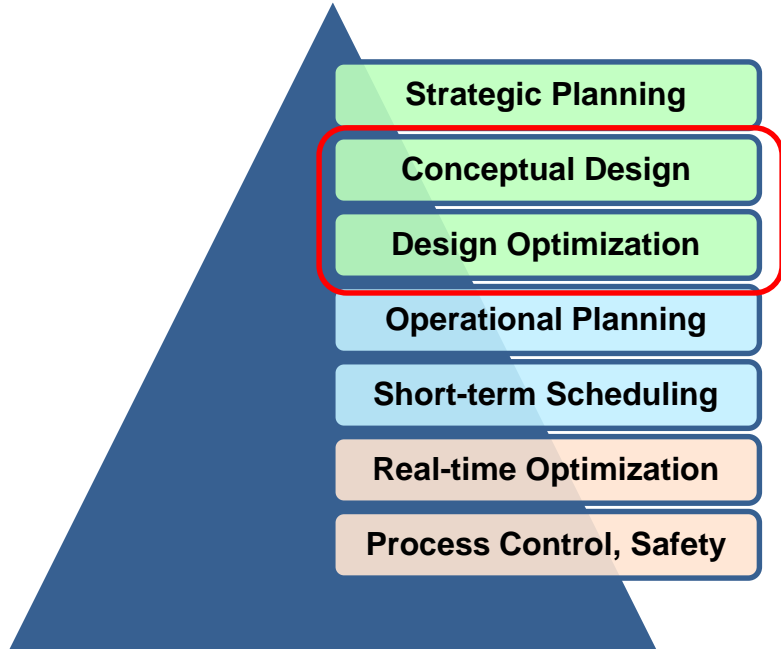


Global CO₂ emissions from industry: key sectors

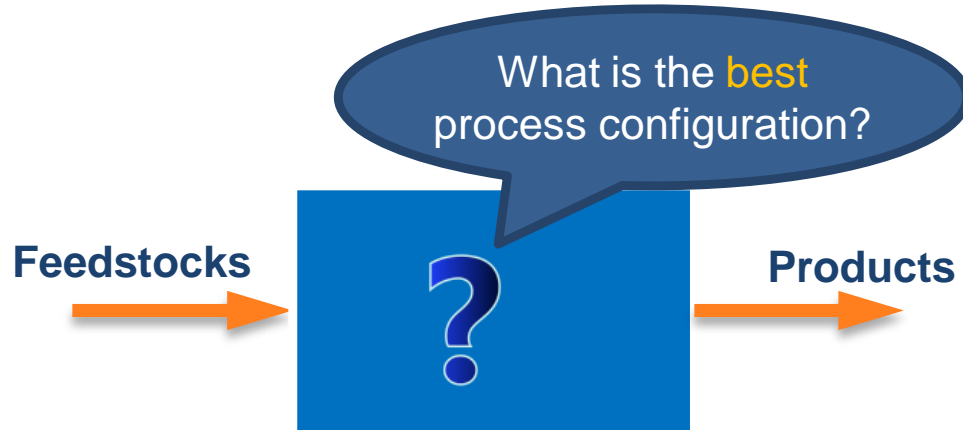


Source: World Economic Forum

Process Systems Engineering Opportunities



Fundamental Questions in Process Design

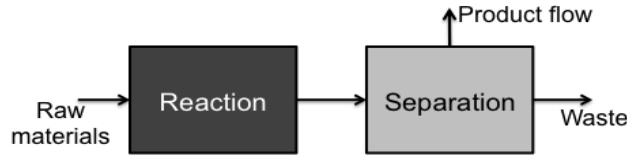


- **Feasibility**
 - product quality, demand
- **Optimality**
 - Cost, energy consumption, waste, emission
 - Profit, NPV, ROI
- Operability, flexibility
- Sustainability, circularity
- Safety

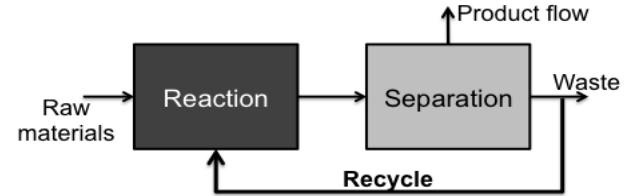
Pillars of Process Design



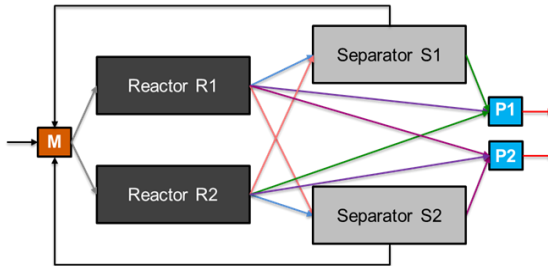
Conceptual Design



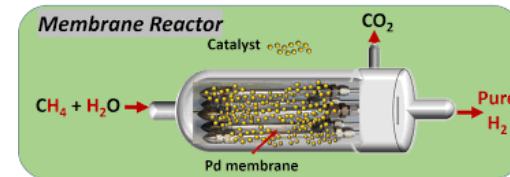
Process Integration



Process Synthesis



Process Intensification

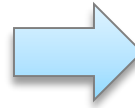
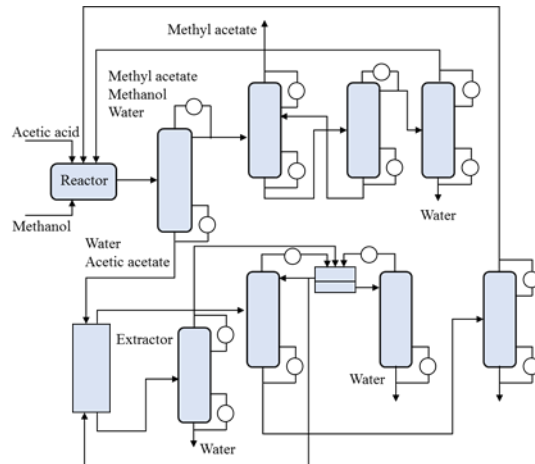


Intensification can be viewed as an extreme case of integration (Baldea, 2015)

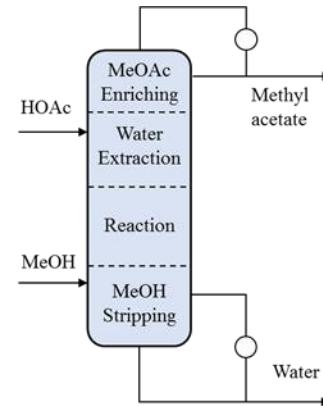
Process Intensification

- Synergistic combination of multiple phenomena and tasks to achieve drastic improvement
- Examples include reactive distillation, membrane reactors, dividing wall columns

Conventional Process



Eastman Chemical Task-integrated Column



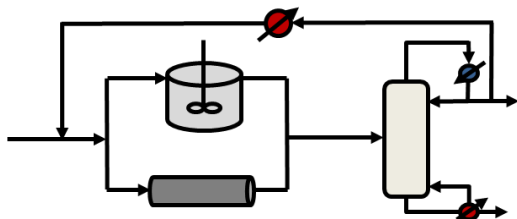
Computer-aided Process Intensification



- Synthesis
 - How can we systematically discover innovative and intensified process systems without waiting for “eureka moments”?
- Identification
 - How can we systematically identify intensification pathways or “intensification hotspots”?
- Analysis
 - Under what conditions is intensification desirable?
 - Can we measure intensifiability? How?
 - Which phenomena should be intensified?

Representation is fundamental to systematic process innovation

Representation Methods



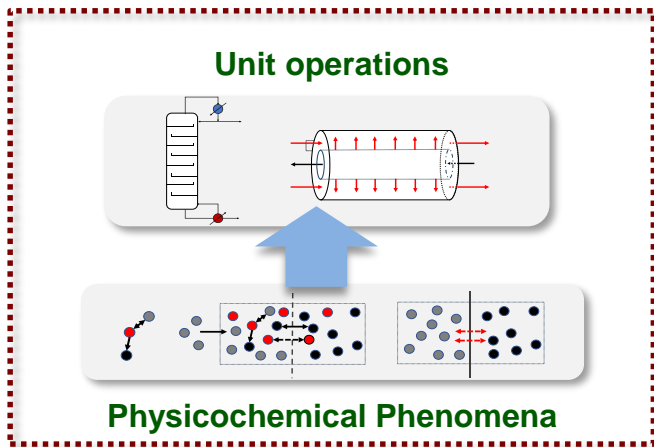
Superstructure Representation

Optimization-based Process Synthesis

Systematic screening of alternatives (Mencareli, Chen, Pagot, Grossmann, *Comput. Chem. Eng.*, 2020, 136, 106808)

- Postulate a process **superstructure** (e.g., P-graph, STN, SEN, UPCS, etc.) and formulate a mathematical program
- Use optimization algorithm to search configurations

Unit operation Focus



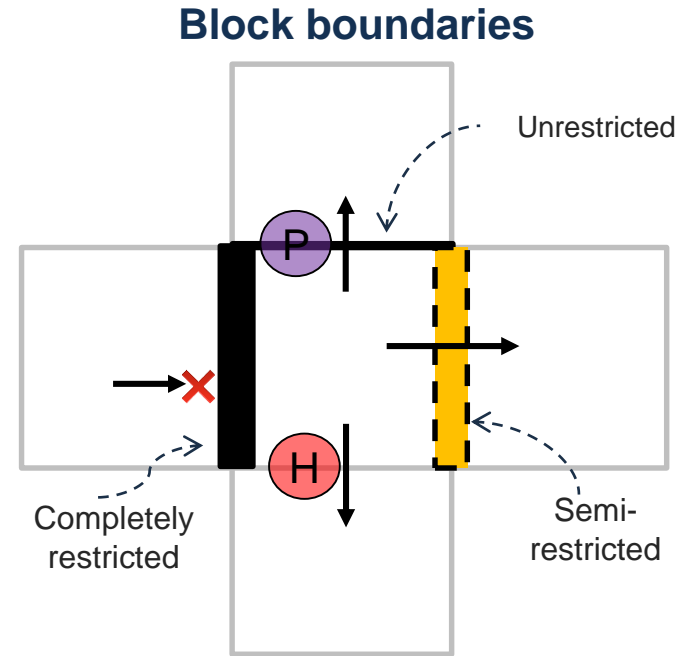
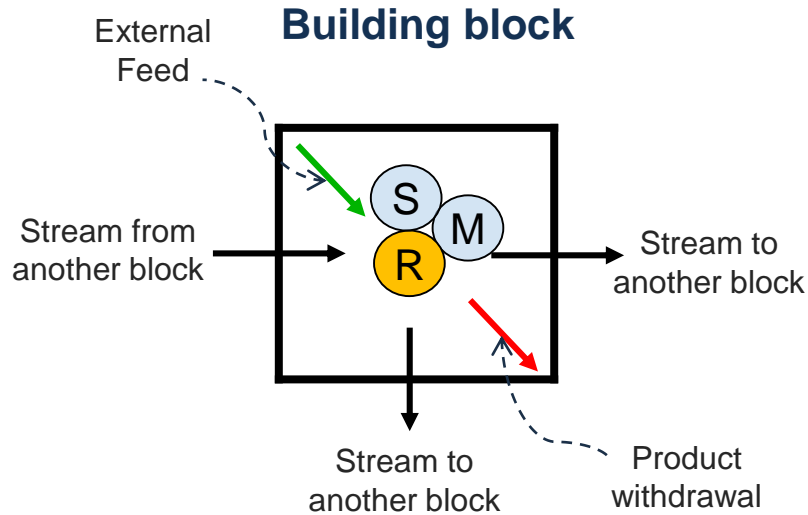
Phenomena/Functional/Modular Approaches

- **Phenomena Building Blocks (PBB)**
 - Lutze et al., 2010; Babi et al., 2014; Tula et al., 2017
- **Generalized Modular Framework (GMF)**
 - Papalexandri and Pistikopoulos, 1996; Tian et al. 2018
- **IDEAS Approach and State-space Representation**
 - Wilson and Manousiouthakis, 2000; Cruz and Manousiouthakis, 2017
- **Functional modules**
 - Freund and Sundmacher, 2008; Kaiser et al., 2018

Departure from Unit Operation Focus

Representation using Design Building Blocks

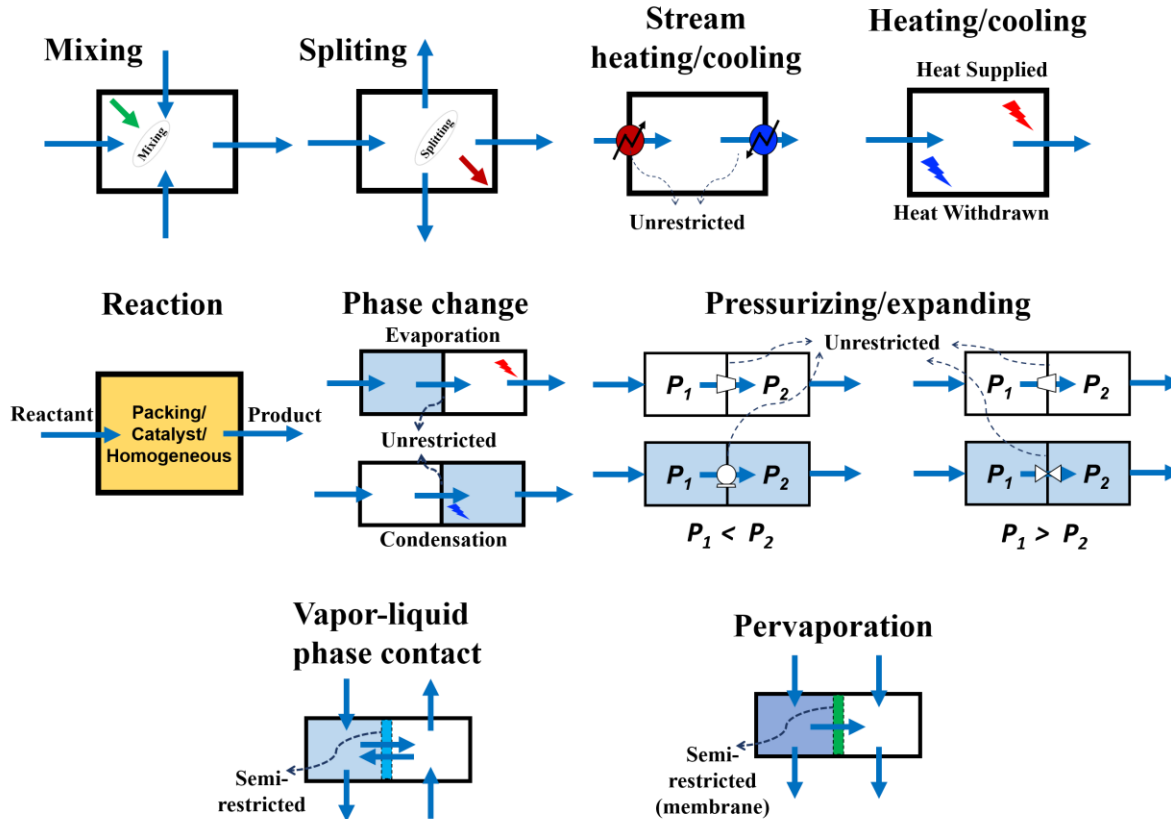
Demirel, Li and Hasan, *Compute. Chem. Eng.*, 2017



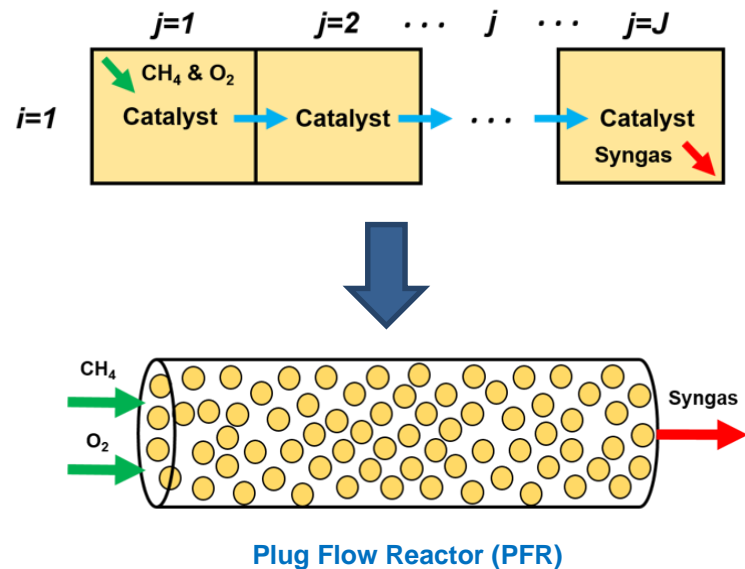
Physical Attributes: Temperature, Pressure, Composition
A unit use of a material (e.g. catalyst)

3 types of Building Block Boundaries

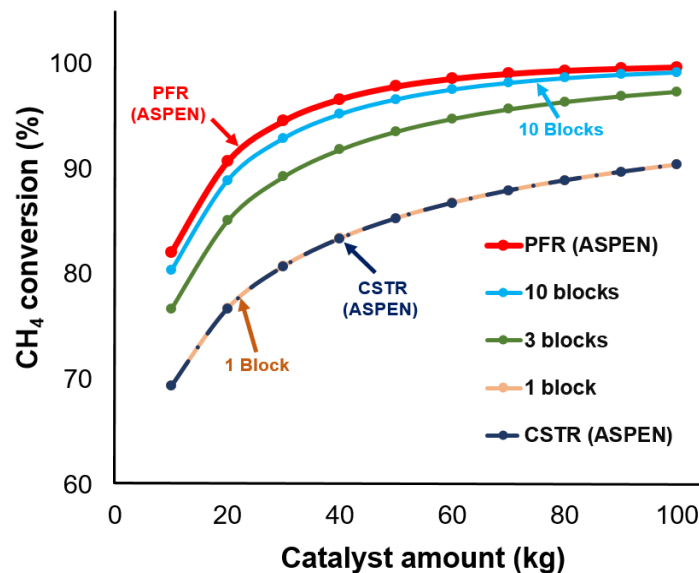
Phenomena Representation



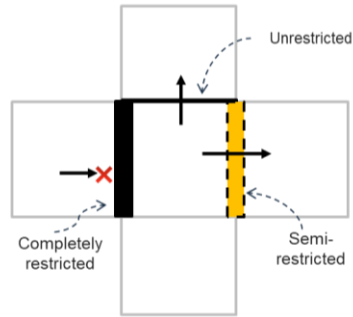
Building Blocks to Phenomena to Unit Operations



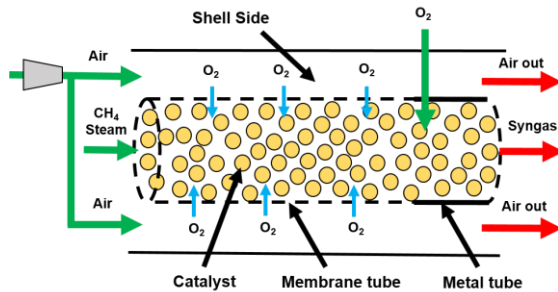
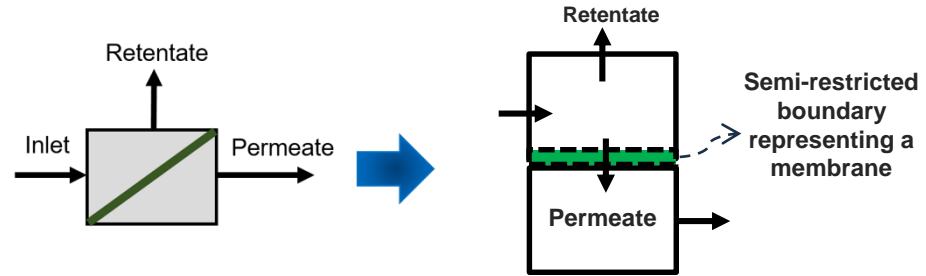
Methane partial oxidation (POX)
(De Groot and Froment, 1996)



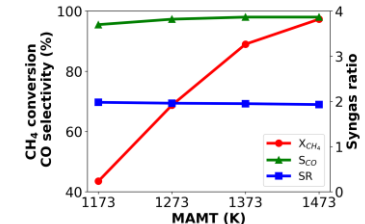
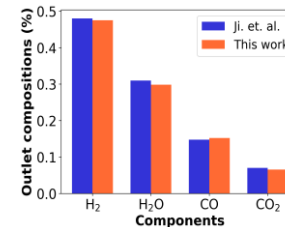
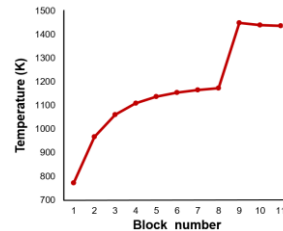
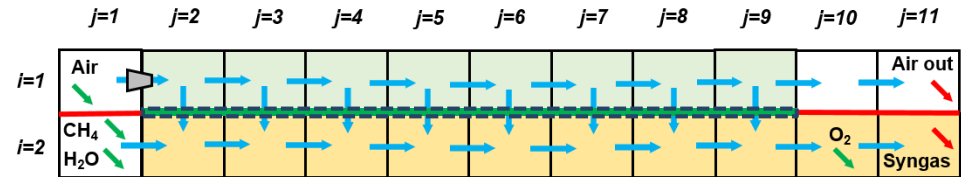
Building Blocks to Phenomena to Unit Operations



Membrane separator

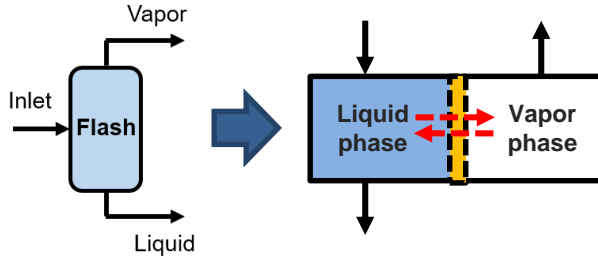


Partially intensified side-feed membrane reactor

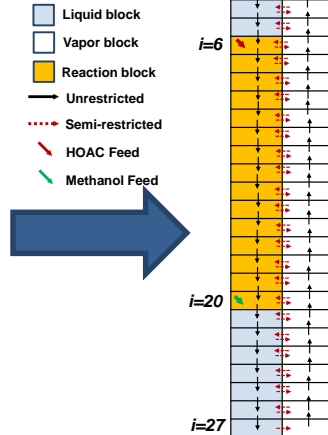
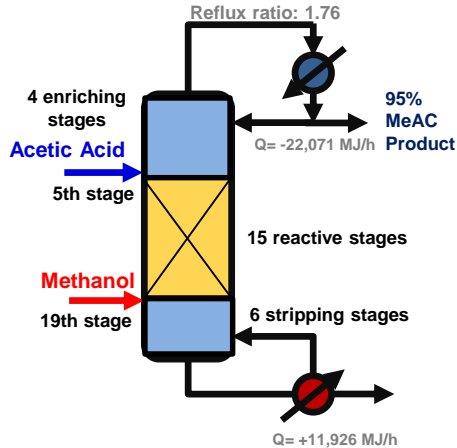


Building Blocks to Phenomena to Unit Operations

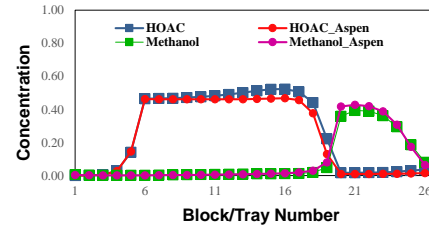
Flash separator



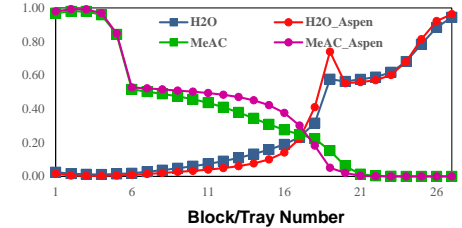
Eastman Chemical task-integrated column



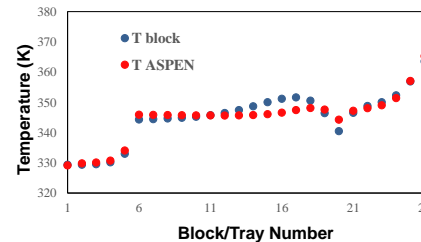
Liquid Mole fraction (Reactants)



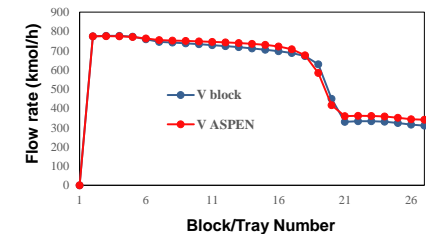
Liquid Mole fraction (Products)



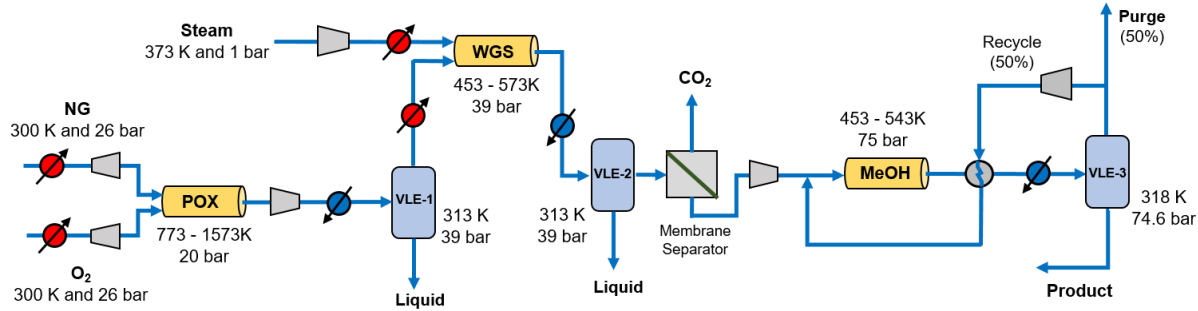
Temperature Profile



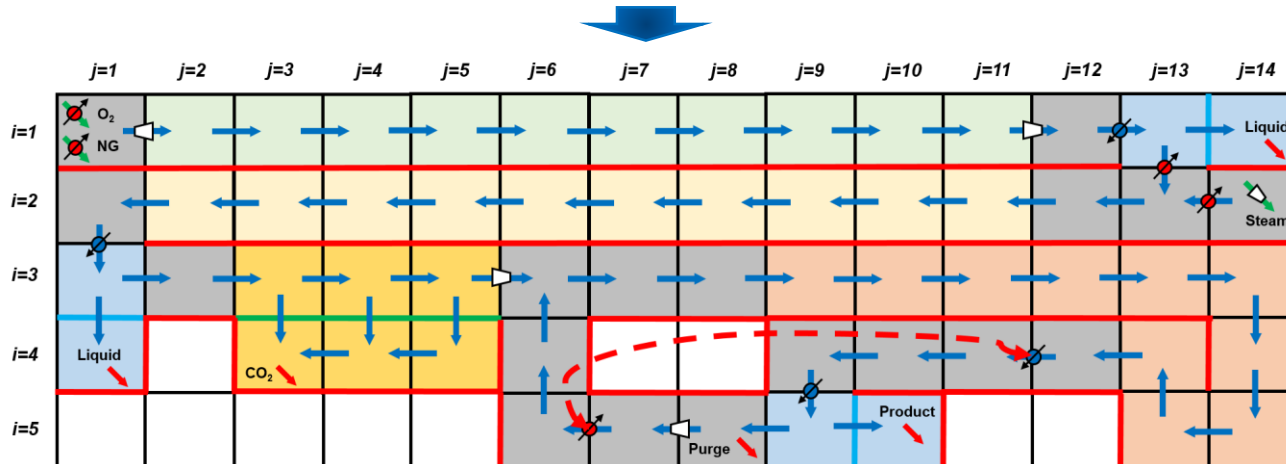
Vapor Flow Rate



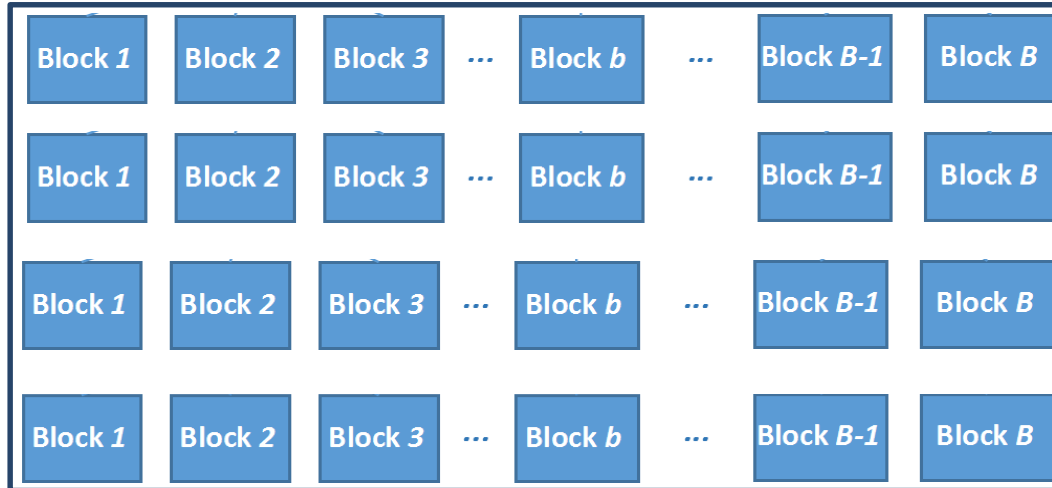
Process Flowsheet Representation



Base Case:
ROI: 38%



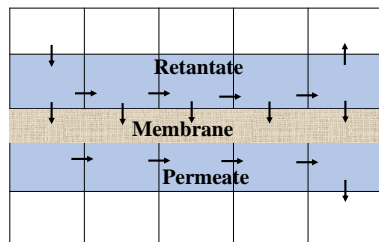
A General Representation for Process Synthesis and Intensification



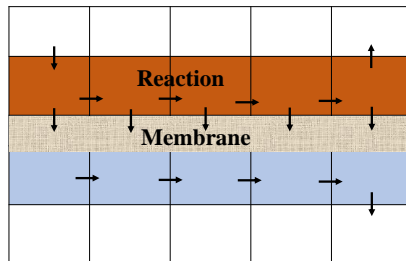
Intensified Equipment Alternatives

Building Block Superstructure

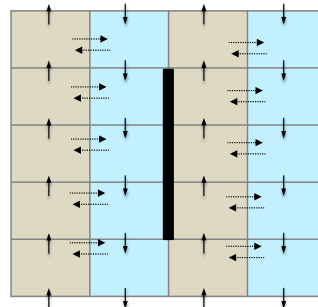
a) Gas Separation Membrane



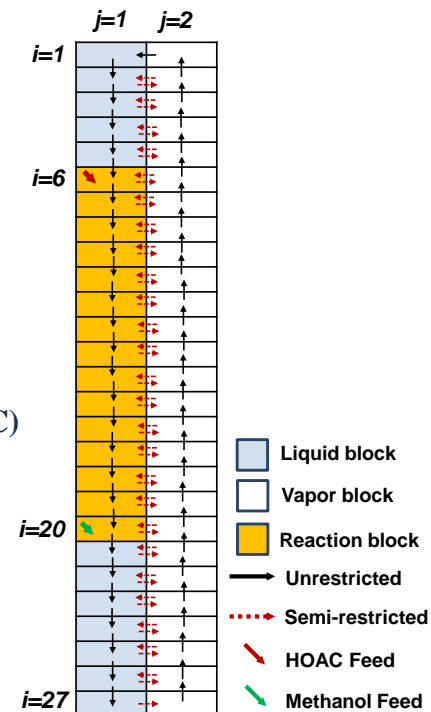
b) Membrane Reactor



c) Dividing Wall Column Distillation (DWC)



d) Eastman Chemical Methyl Acetate Column



- Liquid block
- Vapor block
- Reaction block
- Unrestricted
- Semi-restricted
- HOAC Feed
- Methanol Feed

Mathematical Model Formulation

Given: Process Feed; Product purity and recovery; Material properties; Physicochemical data

Objective: Maximize annual profit, product yield; Minimize cost, utility, total volume, etc.

$$\text{MIN } f(\mathbf{x}, \mathbf{z})$$

s. t.

$$h_t(\mathbf{x}) = 0 \quad \forall t = 1, \dots, T$$

$$g_q(\mathbf{z}) \leq 0 \quad \forall q = 1, \dots, Q$$

$$g_l(\mathbf{x}, \mathbf{z}) \leq 0 \quad \forall q = 1, \dots, L$$

$$\mathbf{x}^L \leq \mathbf{x} \leq \mathbf{x}^U \quad \mathbf{x} \in R^{I \times J \times K}$$

$$\mathbf{z} = \{0, 1\}^S$$

Balance constraints:
Material and Energy balances

Assignment Constraints:
Phenomena and material assignments
and Phase Determination

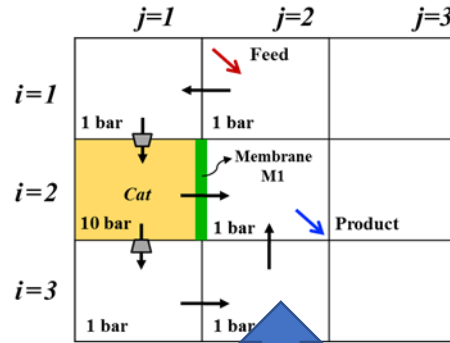
Logical Constraints:
Relating discrete variables with the
continuous variables. These also
includes phenomena models
(rigorous/short-cut/high-level).

\mathbf{x} : Continuous mass and
energy flow variables

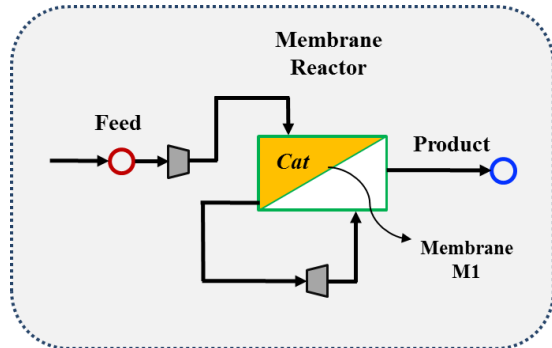
\mathbf{z} : binary variables for
assigning phenomena,
boundaries and functional
materials

Transition from Building Blocks to PFDs

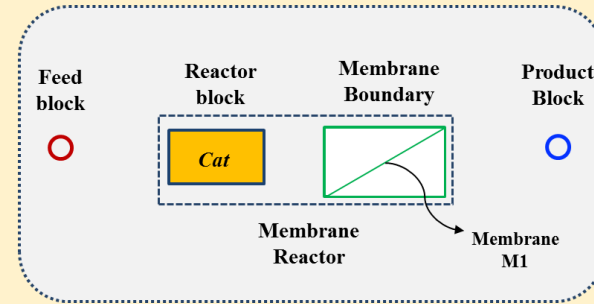
Block Superstructure Result



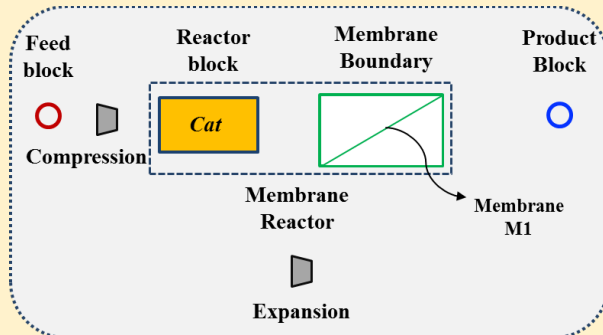
Process Flow Diagram



Identification of Major Operations



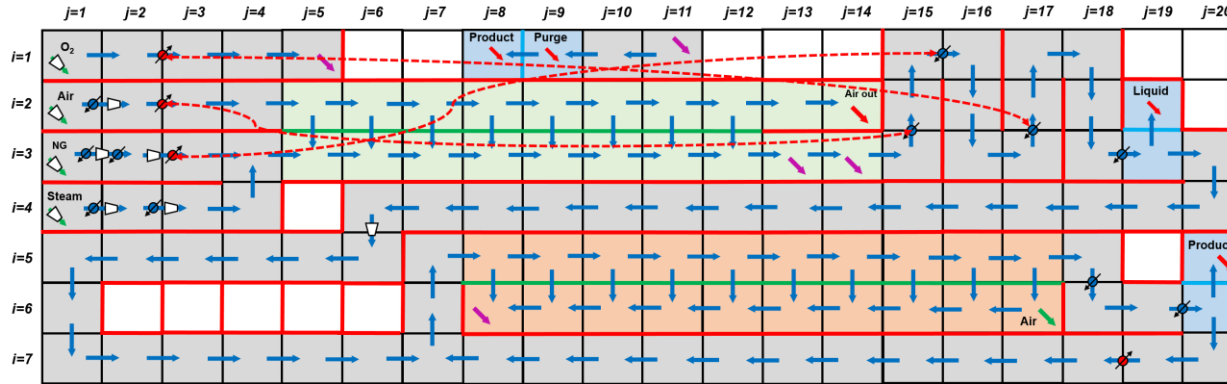
Identification of Auxiliary Operations



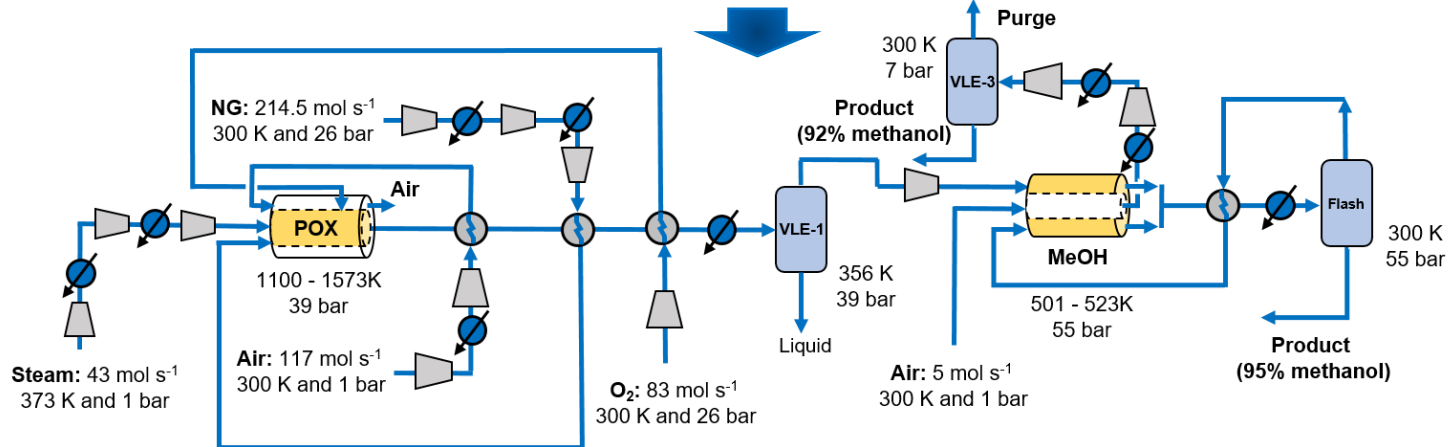
Process Synthesis



Monjur and Hasan, AIChE J., 2022



ROI: 52%



SPICE Framework



User Inputs

- Feedstocks, products and utility specifications (compositions, flow rates, T , P)
- Design targets (purity, product demands, emission, specific considerations, etc.)

Material Library

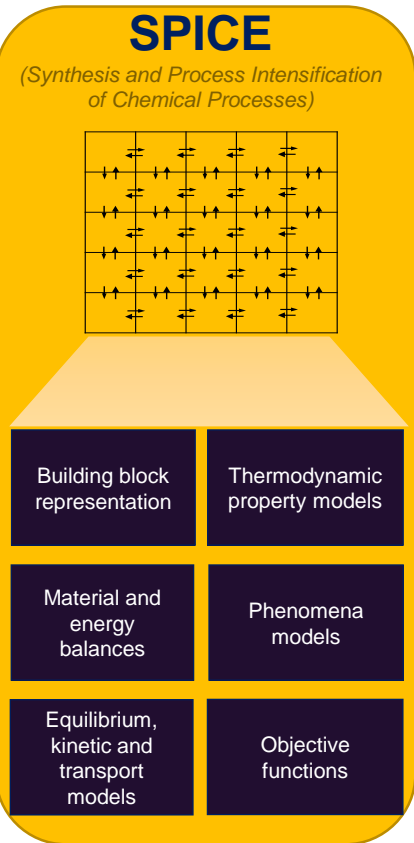
- Reaction library (type, density, porosity, catalyst, kinetics, stoichiometry, selectivity)
- Membrane library (type, permeance, selectivity)
- User defined data, if any

Techno-Economic and Life Cycle Data

- Cost functions and economic parameters (scaling factors, plant life, depreciation, annualizing factor, utility costs, fixed costs, etc.)
- Equipment sizing data
- Emission parameters

Target Representations

- Phenomenal or task level
- Unit operations level
- Flowsheet and process network levels



Targeting, Conceptual Design, and Benchmarking

- Non-intensified equipment
- Intensified equipment

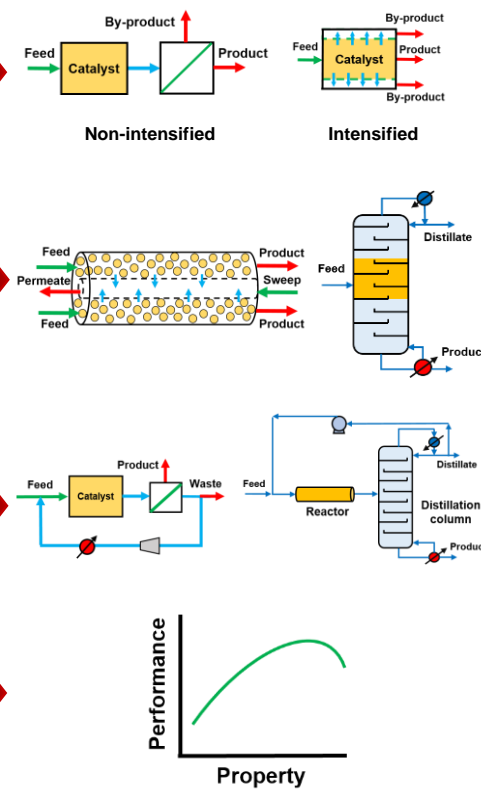
Process Library

- Automated generation of intensified equipment configurations and process flow diagrams (PFDs)
- Simulation, sensitivity analysis, TEA, LCA

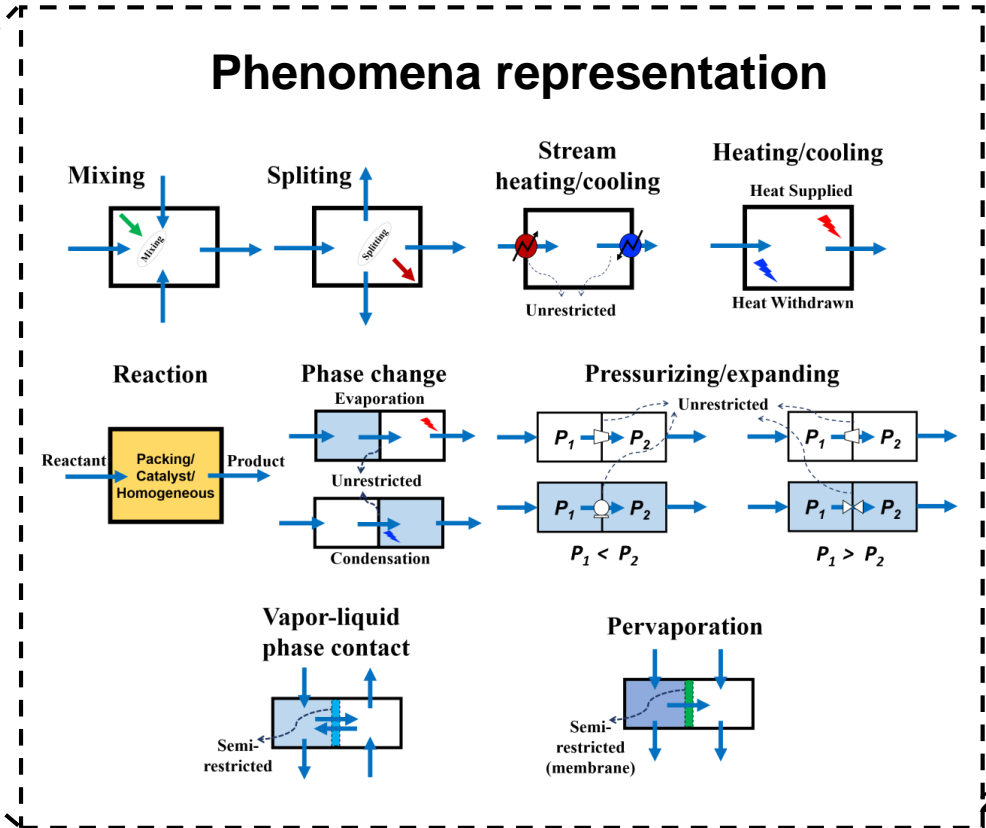
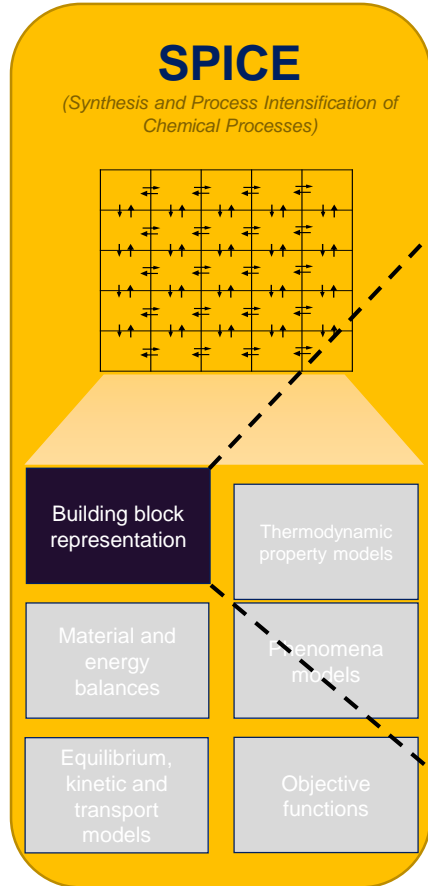
Process Synthesis and Optimization

- Automated screening of process networks
- Design optimization, and sustainability
- Simultaneous heat integration

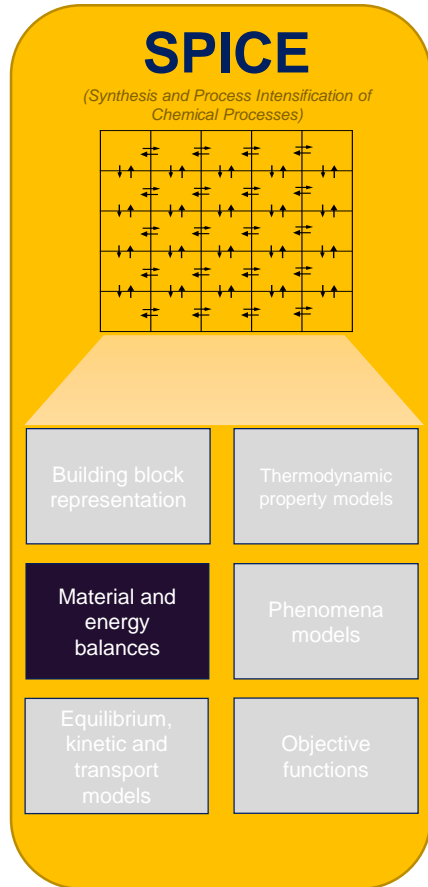
Material-Property-Performance Mapping



Building Block Representation



Model Formulation



Given: Process Feed; Product purity and recovery; Material properties; Physicochemical data

Objective: Maximize annual profit, product yield; Minimize cost, utility, total volume, etc.

$$\text{MIN } f(x, z)$$

s. t.

$$h_t(x) = 0 \quad \forall t = 1, \dots, T$$

$$g_q(z) \leq 0 \quad \forall q = 1, \dots, Q$$

$$g_l(x, z) \leq 0 \quad \forall l = 1, \dots, L$$

$$x^L \leq x \leq x^U \quad x \in R^{I \times J \times K}$$

$$z = \{0, 1\}^S$$

Balance constraints:
Material and Energy balances

Assignment Constraints:
Phenomena and material assignments
and Phase Determination

Logical Constraints:
Relating discrete variables with the
continuous variables. These also
includes phenomena models
(rigorous/short-cut/high-level).

x: Continuous mass and
energy flow variables

z: binary variables for
assigning phenomena,
boundaries and functional
materials

Material balance around each block:

$$F_{i,j-1,k} - F_{i,j,k} + R_{i-1,j,k} - R_{i,j,k} + G_{i,j,k} + \sum_{j \in FS} M_{i,j,k,j} - \sum_{j \in PS} P_{i,j,k,j} = 0$$

Reaction phenomena model:

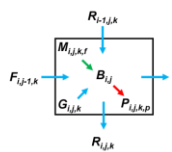
$$G_{i,j,k} = \sum_{r \in RECAT} f(T_{i,j}, P_{i,j}, y_{i,j,k}, k_{i,j,r,e}) \quad k_{i,j,r,e} = f^{TSM}(T_{i,j}, P_{i,j}, y_{i,j,k}, k_{i,j,r,e})$$

Energy balance around each block:

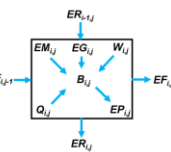
$$EF_{i,j-1} - EF_{i,j} + ER_{i-1,j} - ER_{i,j} + EG_{i,j} + EM_{i,j} - EP_{i,j} + Q_{i,j} + W_{i,j} = 0$$

Separation phenomena model:

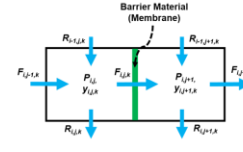
$$F_{i,j,k,1} = f(T_{i,j}, P_{i,j}, y_{i,j,k}, T_{i,j+1}, P_{i,j+1}, y_{i,j+1,k})$$



Material Flow Variables



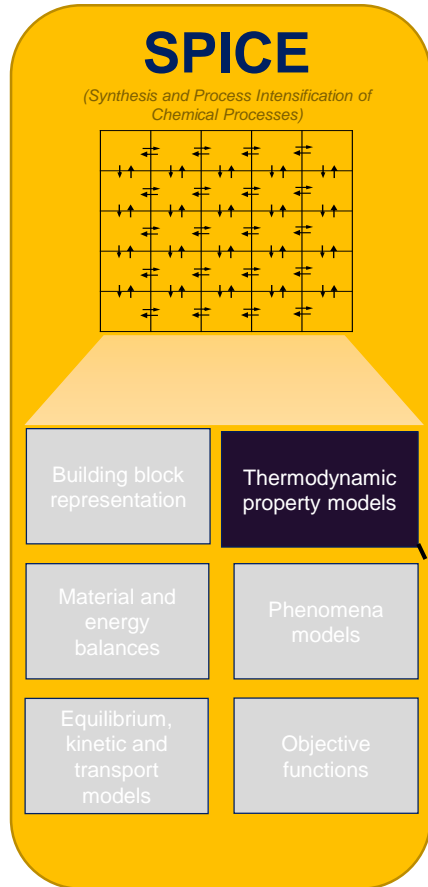
Energy Flow Variables



Minimize total annual cost (TAC)

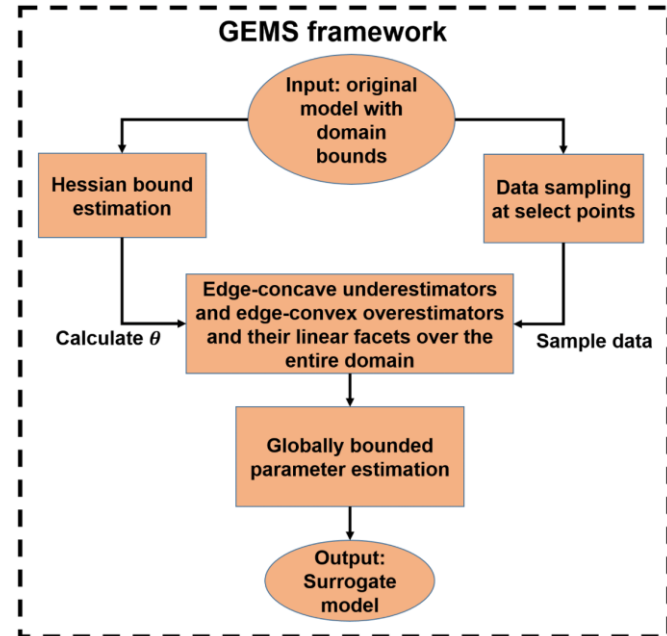
$$\begin{aligned} \text{min TAC} = & \sum_{i \in I} \sum_{j \in J} \sum_{k \in K} UC^I M_{i,j,k,j} + UC^{BL} \sum_{i \in I} \sum_{j \in J} W_{i,j}^{\text{top}} \\ & + UC^{CU} \sum_{i \in I} \sum_{j \in J} Q_{i,j}^C + UC^{HU} \sum_{i \in I} \sum_{j \in J} Q_{i,j}^H \\ & + \gamma_{sc} \sum_{c \in C} C_c \left(\frac{S_c}{SC^{ref}} \right)^{nc} \end{aligned}$$

Thermodynamic Models



GEMS (Guaranteed Error-bounded Surrogate Modeling)

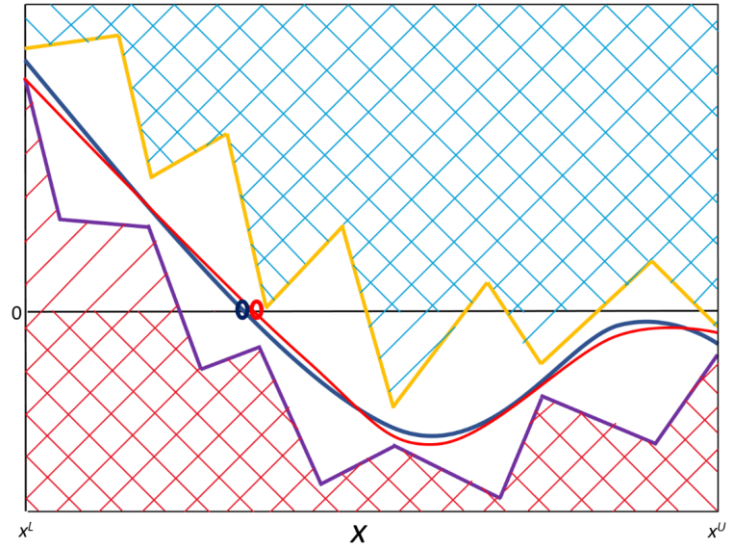
prescribed **error-bounded** outputs, even though based on **model based sampling**



Error Bounded surrogate model parameter estimation: NLP Formulation

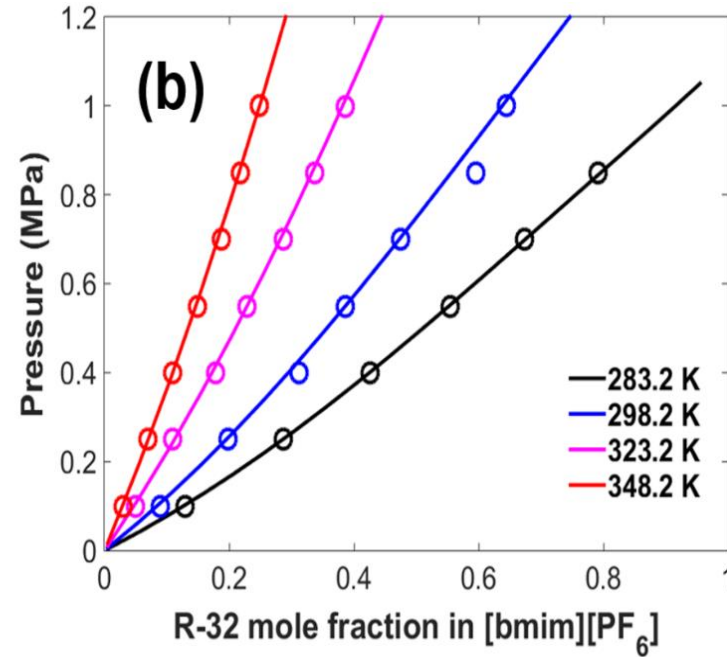
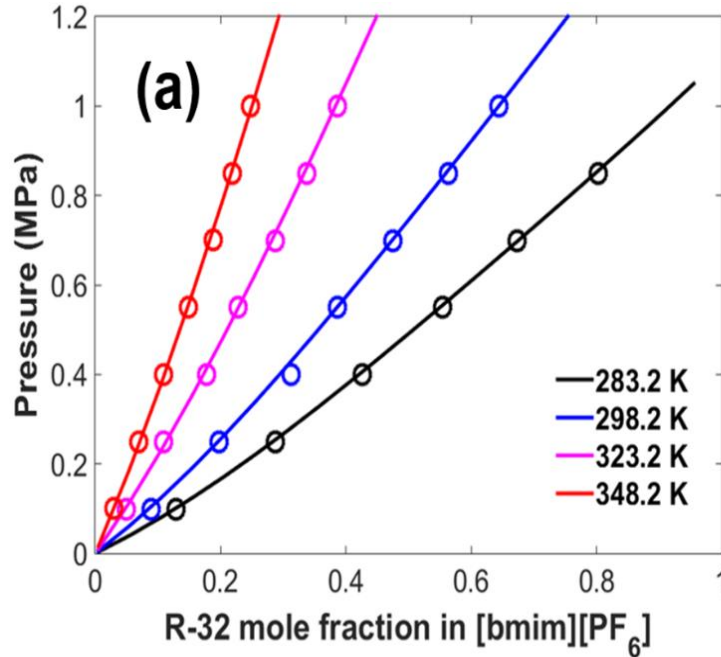


$\min \sum_{n=1}^N \left[\left(\frac{\hat{f}_i - s_i}{\hat{f}_i} \right)^2 + \left(\frac{f_i^U - \hat{f}_i}{\hat{f}_i} \right)^2 + \left(\frac{\hat{f}_i - f_i^L}{\hat{f}_i} \right)^2 \right]$	Objective function
$s_i = \alpha + \sum_{n=1}^N \beta_n \hat{x}_{i,n} + \sum_{n=1}^N \sum_{m=n}^N \gamma_{n,m} \hat{x}_{i,n} \hat{x}_{i,m}$	Surrogate form
$f_i^L - \sum_{n=1}^N \theta_n^{f,L} \cdot (x_{i,n,v} - x_{i,n}^{int})^2 = s_i$ $- \sum_{n=1}^N \theta_n^{s,L} \cdot (x_{i,n,v} - x_{i,n}^{int})^2$ $f_i^U + \sum_{n=1}^N \theta_n^{f,U} \cdot (x_{i,n,v} - x_{i,n}^{int})^2 = s_i$ $+ \sum_{n=1}^N \theta_n^{s,U} \cdot (x_{i,n,v} - x_{i,n}^{int})^2$	Error-bounding of surrogate
$\theta_n^{s,L} \geq \gamma_{n,n}$ $\theta_n^{s,U} \geq -\gamma_{n,n}$	Edge concavity of surrogate underestimators



- $f(x)$
- Feasible region for $L(x)$
- Feasible region for $L(x) \leq 0$
- Actual solution to $f(x) = 0$
- Feasible region for $U(x)$
- Approximate solution to $s(x) = 0$
- Feasible region for $U(x) \geq 0$
- Closest underestimating facets of $f(x)$ shifted by f_i^L
- Closest overestimating facets of $f(x)$ shifted by f_i^U

Solubility prediction

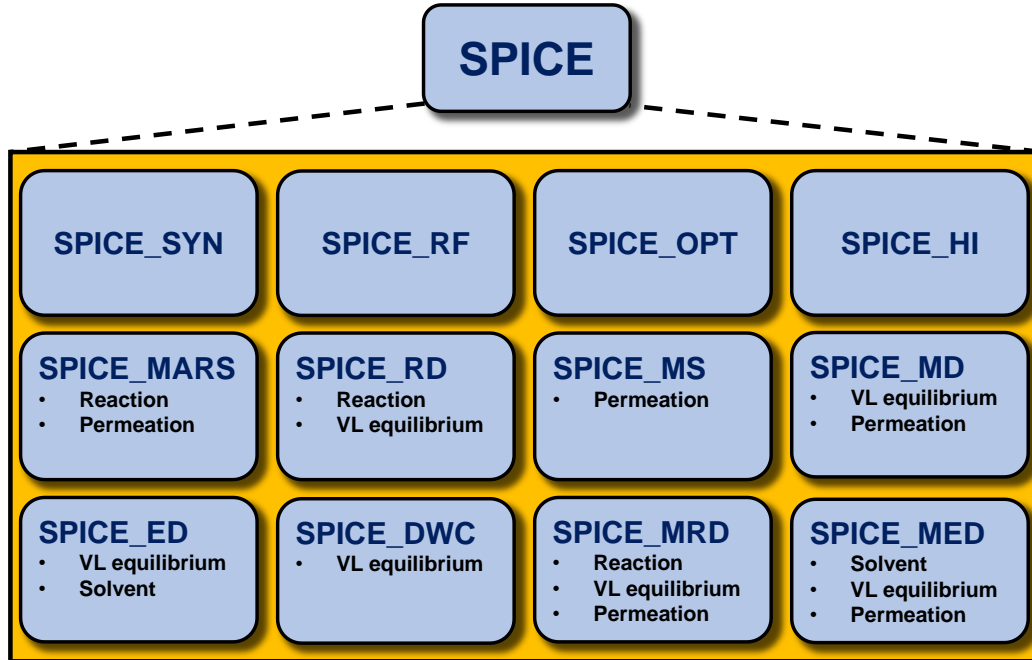


Iftakher et al., In 2022 American Control Conference (ACC), pp. 4814-4819. IEEE, 2022

Solid lines: Gamma-Phi model with (a) Margules activity coefficient, (b) NRTL activity coefficient

Circles: GEMS prediction

SPICE Suits



- Process Synthesis (SPICE_SYN)
- Retrofitting (SPICE_RF)
- Process Optimization (SPICE_OPT)
- Membrane Assisted Reactive Separations (SPICE_MARS)
- Reactive Distillation (SPICE_RD)
- Membrane Separation (SPICE_MS)
- Membrane Distillation (SPICE_MD)
- Extractive Distillation (SPICE_ED)
- Dividing Wall Column (SPICE_DWC)
- ...
- ...

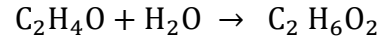
Example: Intensification of Ethylene Glycol Production

Demirel, Li, El-Halwagi and Hasan, ACS Sus. Chem. Eng., 2020

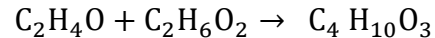
Feed Streams:

- Ethylene Oxide (Sat'd Liquid)
- Water (Liquid at 298 K)

Reaction:



EthyleneOxide + Water → Ethylene Glycol



Ethylene Glycol + Etylene Oxide → Diethylene Glycol

- Liquid phase Homogeneous Reactions

$$r_1 = \exp\left(13.62 - \frac{8220}{T}\right) C_{EO} C_{H_2O}$$

$$r_2 = \exp\left(15.57 - \frac{8700}{T}\right) C_{EO} C_{EG}$$

Target: 25 kmol/h EG production with 95% purity

Semirestricted Boundary (Separation Phenomena)

- Antoine equation for vapor pressure
- Ideal phase equilibrium
- ASPEN Plus parameters for density and vapor pressure

Maximize Return on Investment (ROI)

- Capital Investment
- Raw material Cost
- Hot and Cold Utility Costs

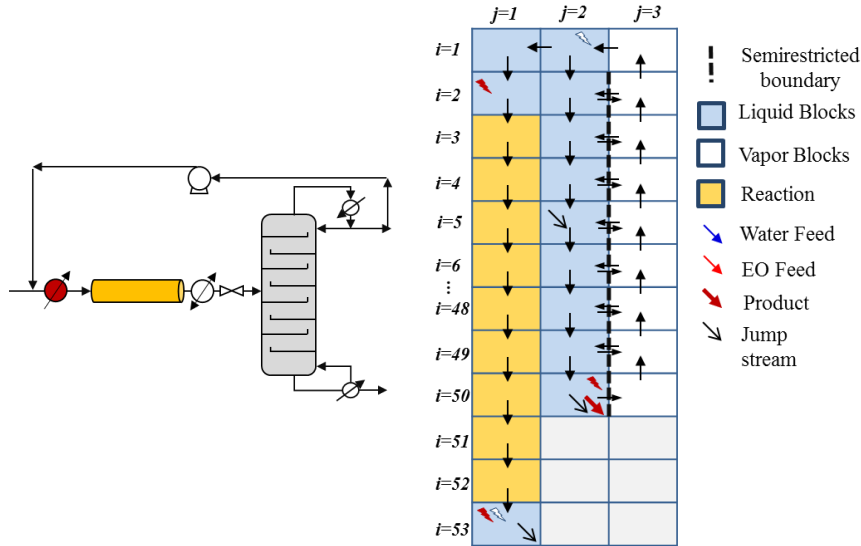
Minimize Total CO₂ emissions:

- Emissions pertained to EO Production
- Electricity
- Steam production

Representation of Base Case Designs



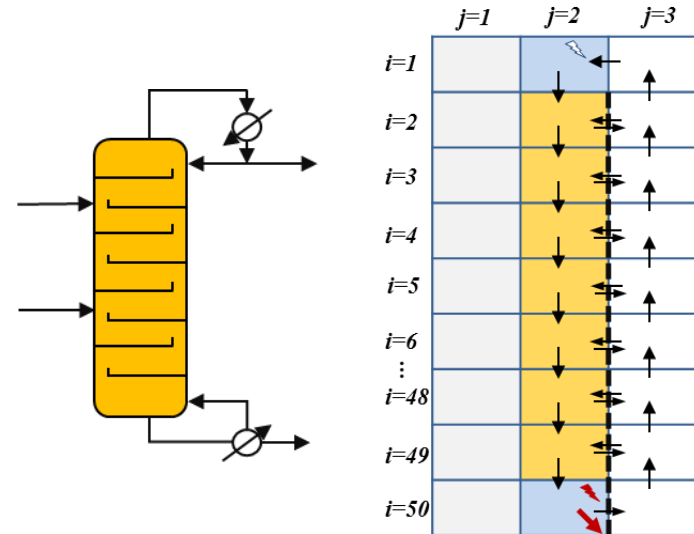
Not intensified: PFR-distillation-recycle



53 × 3 Superstructure:

Plug Flow Reactor (PFR): 50 building blocks (CSTRs-in-series)

Intensified: Single reactive distillation column

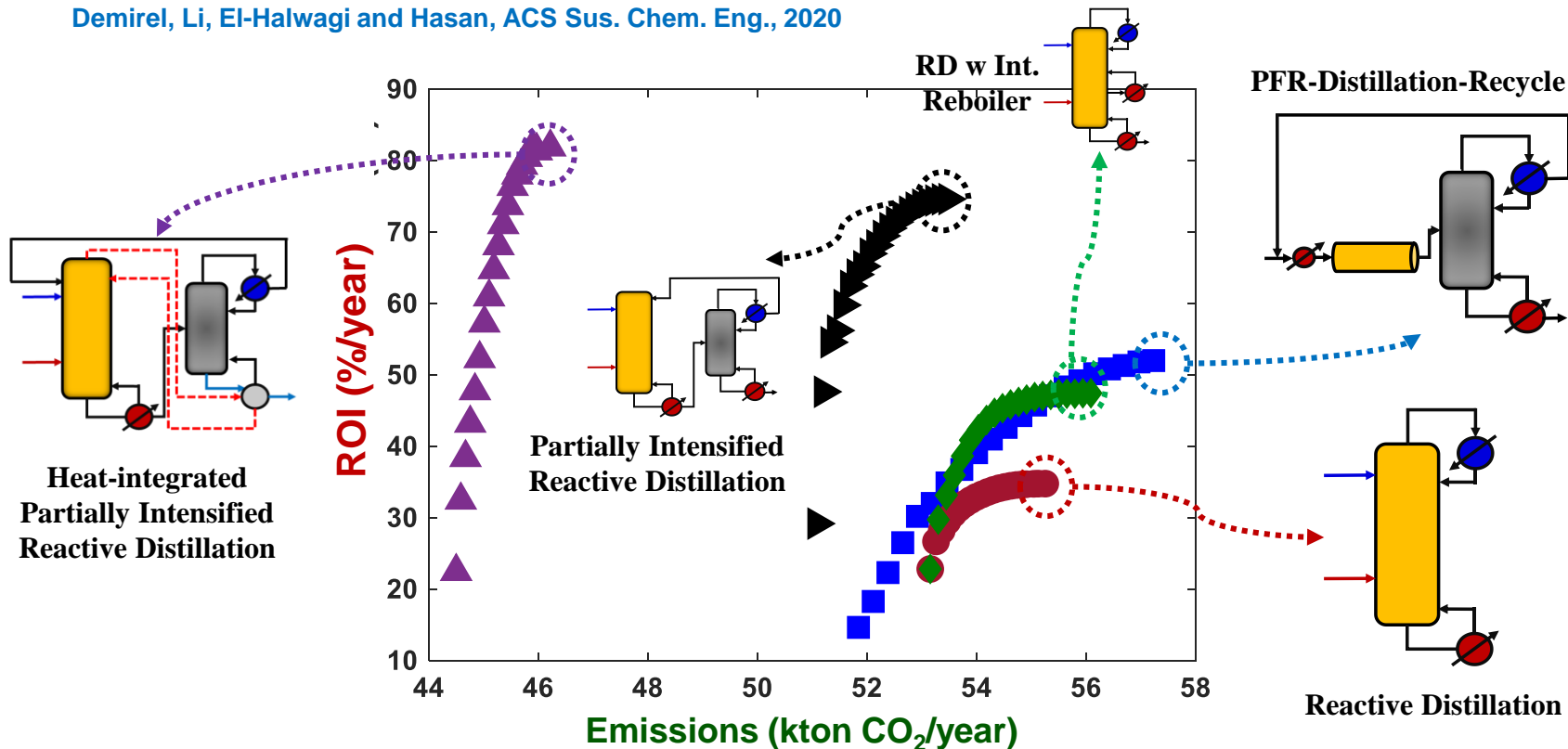


50 × 3 Superstructure:

50 Reactive/non-reactive V-L equilibrium block

Multiobjective Optimization

Demirel, Li, El-Halwagi and Hasan, ACS Sus. Chem. Eng., 2020



Ethylene Glycol (EG) Production



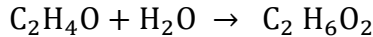
Demirel, Li and Hasan, Ind. Eng. Chem. Res., 2019

Minimize Total Annual Cost for 25 kmol/h EG production with 99.8% purity

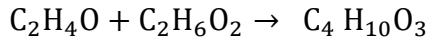
Feeds:

- Ethylene Oxide (Sat'd Liquid at 1 atm)
- Water (Liquid at 298 K)

Reactions:

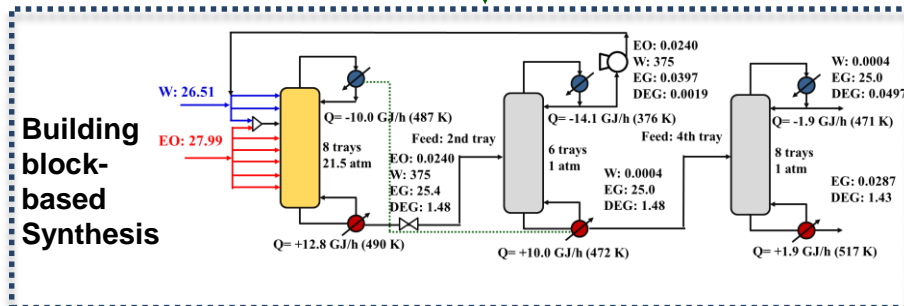
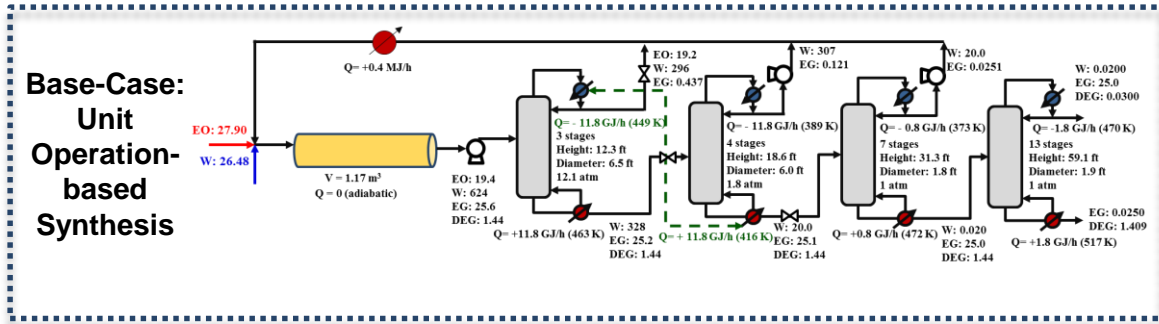


EthyleneOxide + Water → Ethylene Glycol



Ethylene Glycol + Etyhlene Oxide → Diethylene Glycol

- Liquid phase Homogeneous Reaction

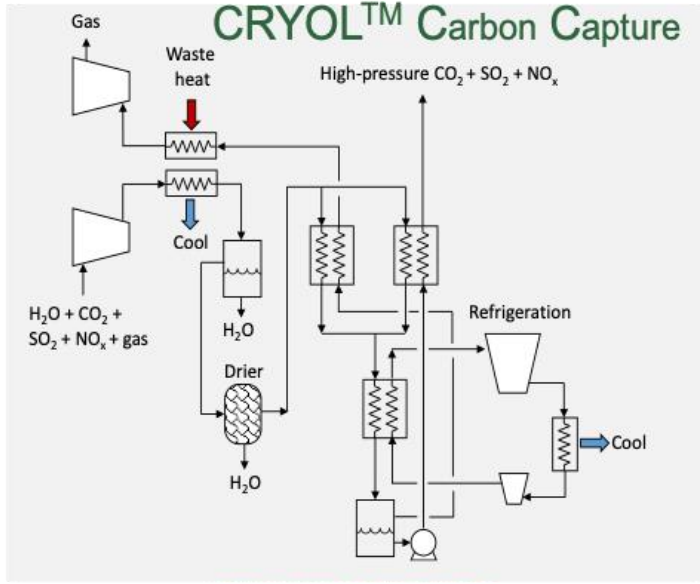


3 major equipment instead of 5 (and 6% less cost)

Post-combustion CO₂ Capture

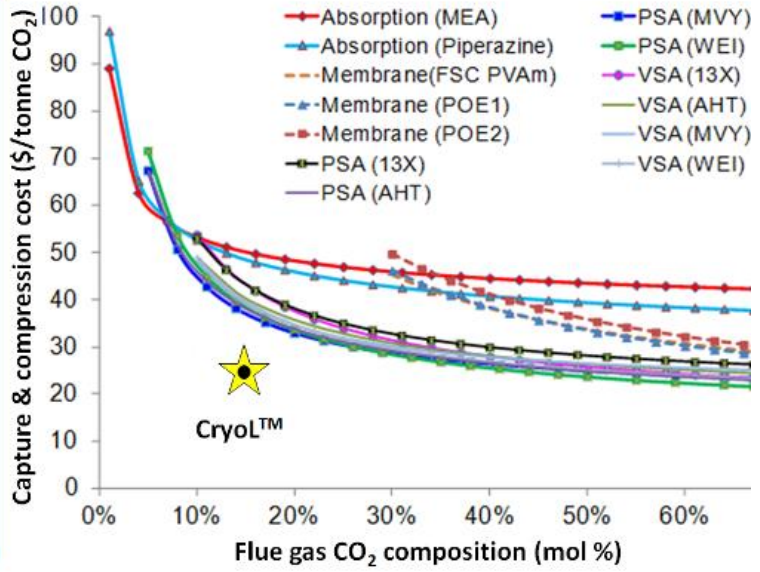


Most cost-effective cryogenic separation process for carbon capture



SIMPLIFIED SCHEMATIC

- ### KEY FEATURES
- Separates >90% CO₂, >99.99% SO₂, and >99.999% NO_x from coal-fired power plant flue gas
 - No sorbents, adsorbents, membranes, or other expensive materials
 - Flexible operation
 - Less than \$25 per tonne CO₂ captured
 - Extensive energy integration and utilization of waste heat
 - Provisional patent filed

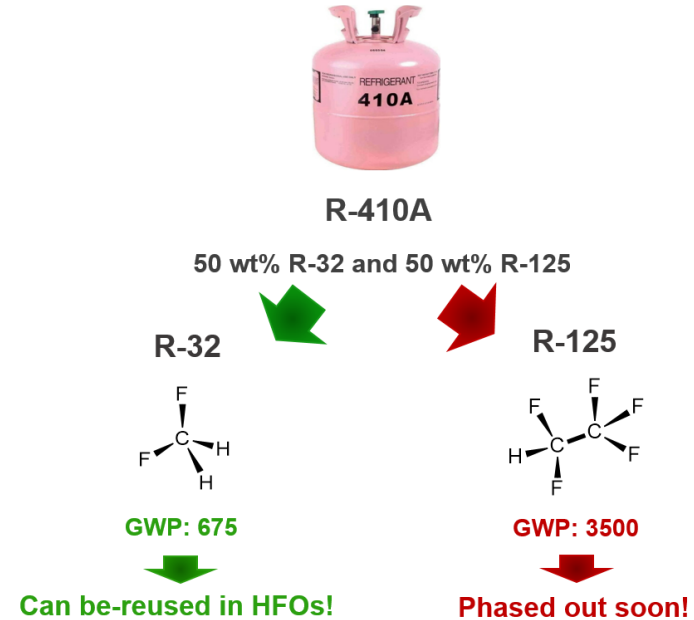


Co-inventors: Sadaf Monjur, Faruque Hasan, Mark Holtzappple

Separation of High-GWP Refrigerant Mixtures



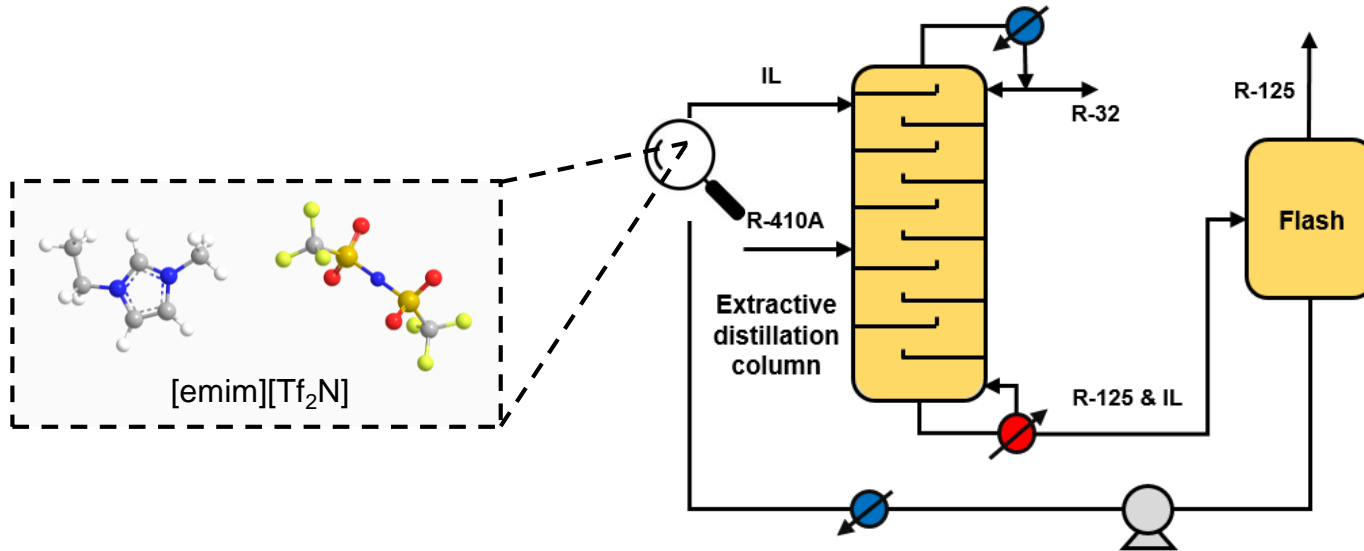
- More than **3 million tons** of refrigerant are in use worldwide
- Hydrofluorocarbons (HFCs)
 - Some have global warming potential (GWP) up to **12,400 CO₂-eq.** (Asensio-Delgado et al., 2021)
 - Leakage of refrigerants contributes to **7.8%** of global GHG emissions (Harders et al., 2022)
- Market value of R-410A separation is over **\$1 billion**
- Challenges in R-410a separation
 - Constituent refrigerants (R-32 and R-125) form **azeotropic mixture**
 - Conventional distillation-based separation becomes highly **energy-** and **cost-intensive**
 - Currently there exist **no commercial** separation techniques for refrigerant mixtures separation



Extractive Distillation-based Separation



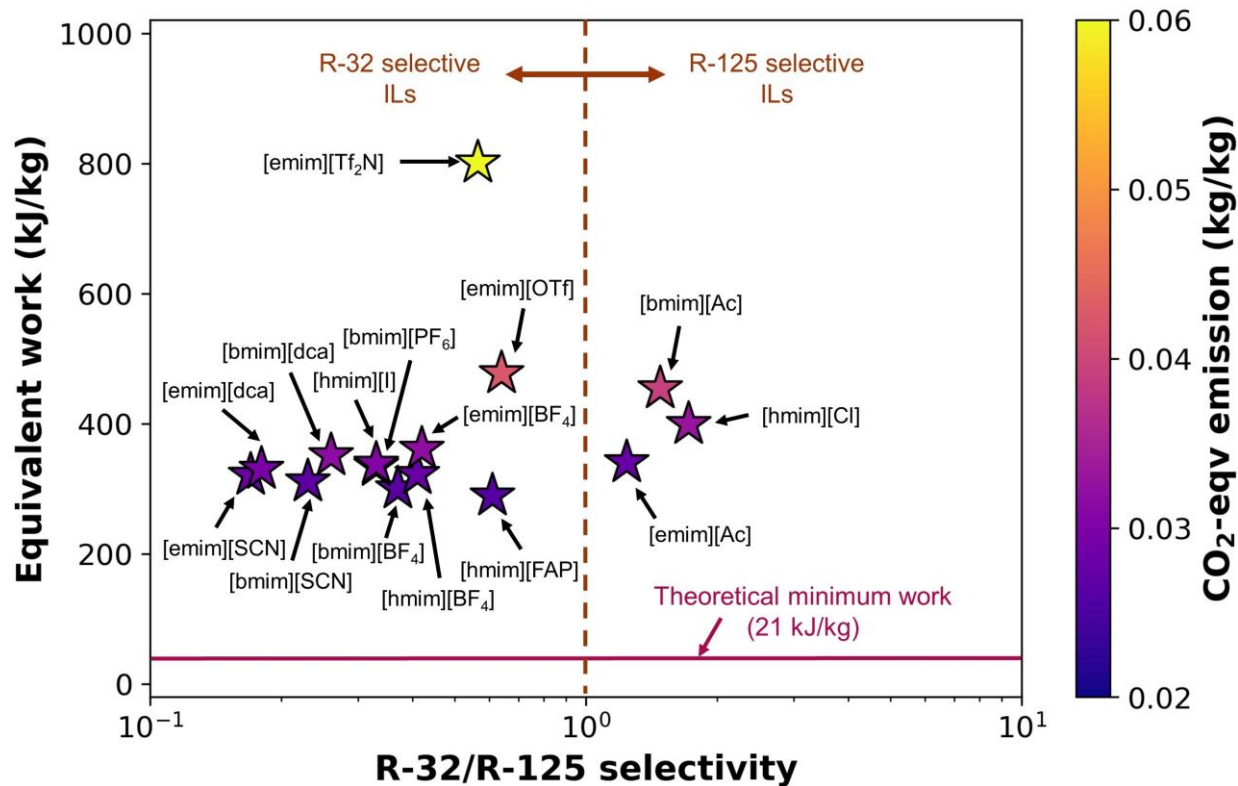
Extractive Distillation (ED) is an **intensification** technique that utilizes a high boiling point **solvent** to change the component **relative volatilities**



PROJECT
EARTH



Process-based Screening of Existing Ionic Liquids



Concluding Remarks

- Process intensification shows potential for addressing some of the new and old challenges in the chemical process industry
- Representation is critical in systematic innovation in process intensification
- Design building blocks
 - Provide a seamless transition between phenomena, tasks, equipment and flowsheets
 - Allow optimization-based approach for automated flowsheet generation, process synthesis and intensification
 - Provide an initial starting point for approaching the central tradeoff in superstructure-based synthesis (between generality and tractability)
- Several outstanding challenges remain to be addressed
 - Generality: It is an open question how many rows and columns should be postulated in the initial superstructure
 - Tractability: New solution techniques are required considering inherent symmetry and degeneracy

Acknowledgement



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Sustainable Process Synthesis and Intensification for Industrial Decarbonization

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Energy Institute



Texas A&M Engineering
Experiment Station

2022 Seminar for Energy Systems Initiative (ESI) at the Center for Advanced Process Decision-Making (CAPD)
November 09, 2022