Synthesis of Optimal PSA Cycles for Hydrogen/CO<sub>2</sub> Separation

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## Introduction



- Shifted syngas composition: H<sub>2</sub>, CO<sub>2</sub> and small amounts of CO, CH<sub>4</sub>, N<sub>2</sub> and Ar
- Conditions: ambient temperature (310 K) and high pressure (7 bar)
- Pre-combustion CO<sub>2</sub> capture technologies: absorption, membranes, adsorption etc.

**Pressure Swing Adsorption** 

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\*http://www.futuregenalliance.org/publications.stm

# What PSA cycle for CO<sub>2</sub> capture?



Light reflux

pressurization

Heavy product pressurization

Heavy reflux

- Numerous PSA configurations suggested in the literature
- No systematic methodology available



## **PSA Superstructure**



- Most PSA cycles in literature have only two cycles interacting at a given time
- Different operating steps realized by varying the control variables  $\alpha(t)$ ,  $\beta(t)$ ,  $\Phi(t)$ ,  $P_{ads}(t)$  and  $P_{des}(t)$



### Cyclic realization from superstructure



## **PSA Model**

Partial differential and algebraic equations (PDAEs)

Component mass balance

$$\epsilon_B \frac{\partial \rho_i}{\partial t} + \rho_s \frac{\partial n_i}{\partial t} + \frac{\partial (\rho_i \nu)}{\partial z} = 0 \qquad i = 1, 2, ..., N_c$$

Linear driving-force equation

$$\frac{\partial n_i}{\partial t} = \frac{k_i}{R_{PV}T} (P_i - P_i^*) \qquad i = 1, 2, \dots, N_c$$

Energy balance

$$\left(\epsilon_B \sum_{i=1}^{N_c} \rho_i (c_p^i - R_E) + \rho_s c_s + \rho_s \sum_{i=1}^{N_c} n_i c_p^i \right) \frac{\partial T}{\partial t} - \rho_s \sum_{i=1}^{N_c} q_i \frac{\partial n_i}{\partial t} + \frac{\partial (\nu h)}{\partial z} = 0$$

Steady-state momentum balance (Ergun equation)

$$-\frac{\partial P}{\partial z} = 150 \frac{\mu\nu}{d_p^2} \frac{(1-\epsilon_B)^2}{\epsilon_B^3} + 1.75 \frac{\rho M\nu^2}{d_p} \frac{1-\epsilon_B}{\epsilon_B^3}$$

Dual-site Langmuir adsorption isotherm  

$$n_{i} = m_{1} \frac{b_{i}P_{i}^{*}}{1 + \sum_{i=1}^{N_{c}} b_{j}P_{j}^{*}} + m_{2} \frac{b_{i}P_{i}^{*}}{1 + \sum_{i=1}^{N_{c}} b_{j}P_{j}^{*}} \quad i = 1, 2, ..., N_{c}$$
Ideal gas law  

$$\rho = \frac{P}{R_{PV}T}$$
Enthalpy  

$$h = \sum_{i=1}^{N_{c}} \rho_{i}(aT + bT^{2} + cT^{3} + dT^{4})$$
Molecular weight  

$$M = \sum_{i=1}^{N_{c}} y_{i}M_{0,i} \quad \text{where} \quad y_{i} = \rho_{i}/\rho$$
Fluid viscosity  

$$\mu = \sum_{i=1}^{N_{c}} \mu_{i}y_{i} \quad \text{where} \quad \mu_{i} = \mu_{0,i} + \mu_{1,i}T$$



# **Optimization Problem**



#### Handling PDAEs: Simultaneous method

#### Sequential method



#### Case Study: Simultaneous Approach

min power c(z) = 0s.t.  $CO_2$  purity  $\geq 0.9$  $CO_2$  recovery > 0.92 $Q_{feed,L} + Q_{feed,H} \ge 35 \text{ kgmol m}^{-2} \text{hr}^{-1}$  $P_{ads} \ge P_d$  $P_{des} \leq P_{feed}$  $P_a \geq P_{feed}$  $0 \leq \alpha(\mathbf{t}_i), \beta(\mathbf{t}_i), \phi(\mathbf{t}_i) \leq 1$  $\forall t_i$  $10 \sec < T_c < 500 \sec$ 101.32 kPa  $\leq P_{ads}(t_i) \leq 1000$  kPa  $\forall t_i$  $P_{des}(t_i) > 50 \text{ kPa}$  $\forall t_i$ 

- Feed composition: 55% H<sub>2</sub> and 45% CO<sub>2</sub>
- Activated carbon is used as the adsorbent
- No. of temporal finite elements: 10
- No. of spatial finite volumes: 10
- Large-scale nonlinear program: solved using AMPL/IPOPT





#### Case Study 2: Simultaneous Approach (contd..)

No. of variables: 10512

No. of constraints: 10434

Optimal cycle time: 424.74s

Feed flux: 96.61 kgmol/(m<sup>2</sup> h)

Power consumption: 46.82 kW h/tonne CO<sub>2</sub>

CPU time ~ 1h



Spatial finite volumes	Full discretization	MATLAB simulation
H <sub>2</sub> purity	93.33%	94.92%
H <sub>2</sub> recovery	91.64%	91.05%
CO <sub>2</sub> purity	90%	89.42%
CO <sub>2</sub> recovery	92%	93.67%



# Summary

- PSA superstructure to determine optimal cycle configurations and design parameters for precombustion CO<sub>2</sub> capture is presented
- Formulated as an optimal control problem with binary feed mixture which is solved using a complete discretization approach
- Optimal PSA cycles obtained for pre-combustion CO<sub>2</sub> capture
- Framework can be used to evaluate different feedstocks and adsorbents



# **Ongoing Work**



# **Ongoing Work**

Limitations of Simultaneous Method:

- Discretization error caused due to fewer temporal and spatial nodes
- Problem size grows cubically with increase in number of spatial and temporal nodes
- Multi-component feed mixtures and multi-adsorbent layers becomes very hard to solve
- Higher CPU times

### **Sequential Method**



#### **Sequential Method**

#### Discretize only in the spatial domain



- NLP problem is decoupled from the embedded dynamic system
- Use of error-controlled integrators like CVODES, DASPK etc.
- Smaller NLP sub-problem



#### **Sequential Method**

- Periodic Operation results in dense constraint Jacobian
- Dominant computational cost: forming and factorization of Jacobian
- Instead work with exact Jacobian-vector and Hessianvector products
- Developed and implemented a novel NLP algorithm to work with exact matrix-vector products and inexact Jacobian
- Tested on small scale Simulated Moving Bed and PSA applications
- Obtained an order of magnitude reduction in CPU time compared to current methods
- Next step: Solve the PSA superstructure optimization problem using the sequential method



## Thank you



# Summary

- PSA superstructure to determine optimal cycle configurations and design parameters for pre-combustion CO<sub>2</sub> capture is presented
- Formulated as an optimal control problem with binary feed mixture which is solved using a complete discretization approach
- Min. power case study resulted in a 2-bed 10step cycle
- Resulted in high purity and recoveries of H<sub>2</sub> and CO<sub>2</sub> with a significantly low power consumption of 46.82 kWh/tonne of CO<sub>2</sub>



#### Case Study: Simultaneous Approach

max

s.t.

- $\begin{array}{ll} \mathrm{CO}_2 \ \mathrm{recovery} \\ c(z) = 0 \\ \mathrm{H}_2 \ \mathrm{purity} \geq 0.9 \\ \mathrm{CO}_2 \ \mathrm{purity} \geq 0.9 \\ Q_{feed,L} + Q_{feed,H} \geq 35 \ \mathrm{kgmol} \ \mathrm{m}^{-2} \mathrm{hr}^{-1} \\ P_{ads} \geq P_d \\ P_{des} \leq P_{feed} \\ P_a \geq P_{feed} \\ 0 \leq \alpha(\mathrm{t}_i), \beta(\mathrm{t}_i), \phi(\mathrm{t}_i) \leq 1 \\ 10 \ \mathrm{sec} \leq T_c \leq 500 \ \mathrm{sec} \\ 101.32 \ \mathrm{kPa} \leq P_{ads}(\mathrm{t}_i) \leq 1000 \ \mathrm{kPa} \\ P_{des}(\mathrm{t}_i) \geq 50 \ \mathrm{kPa} \\ \end{array}$
- Feed composition: 55% H<sub>2</sub> and 45% CO<sub>2</sub>
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#### Case Study: Simultaneous Approach (contd..)

No. of variables: 10512

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CPU time ~ 1h

Spatial finite	Full	MATLAB
volumes	discretization	simulation
H <sub>2</sub> purity	98.20%	95.92%
H <sub>2</sub> recovery	91.09%	91.73%
CO <sub>2</sub> purity	90%	90.99%
CO <sub>2</sub> recovery	97.95%	96.03%