

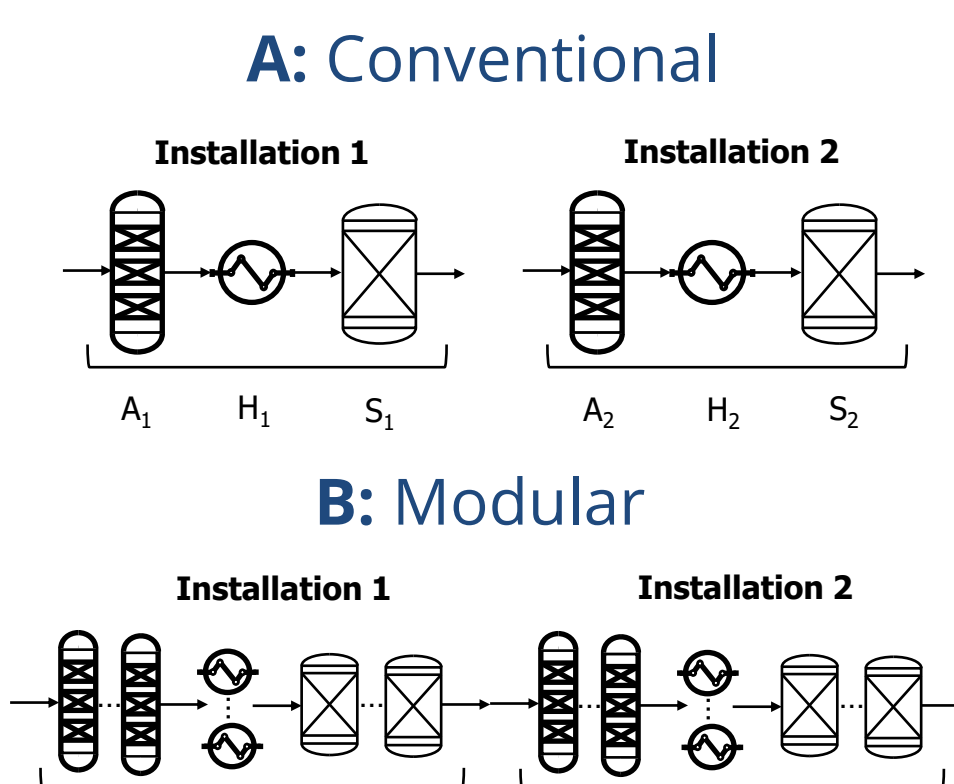
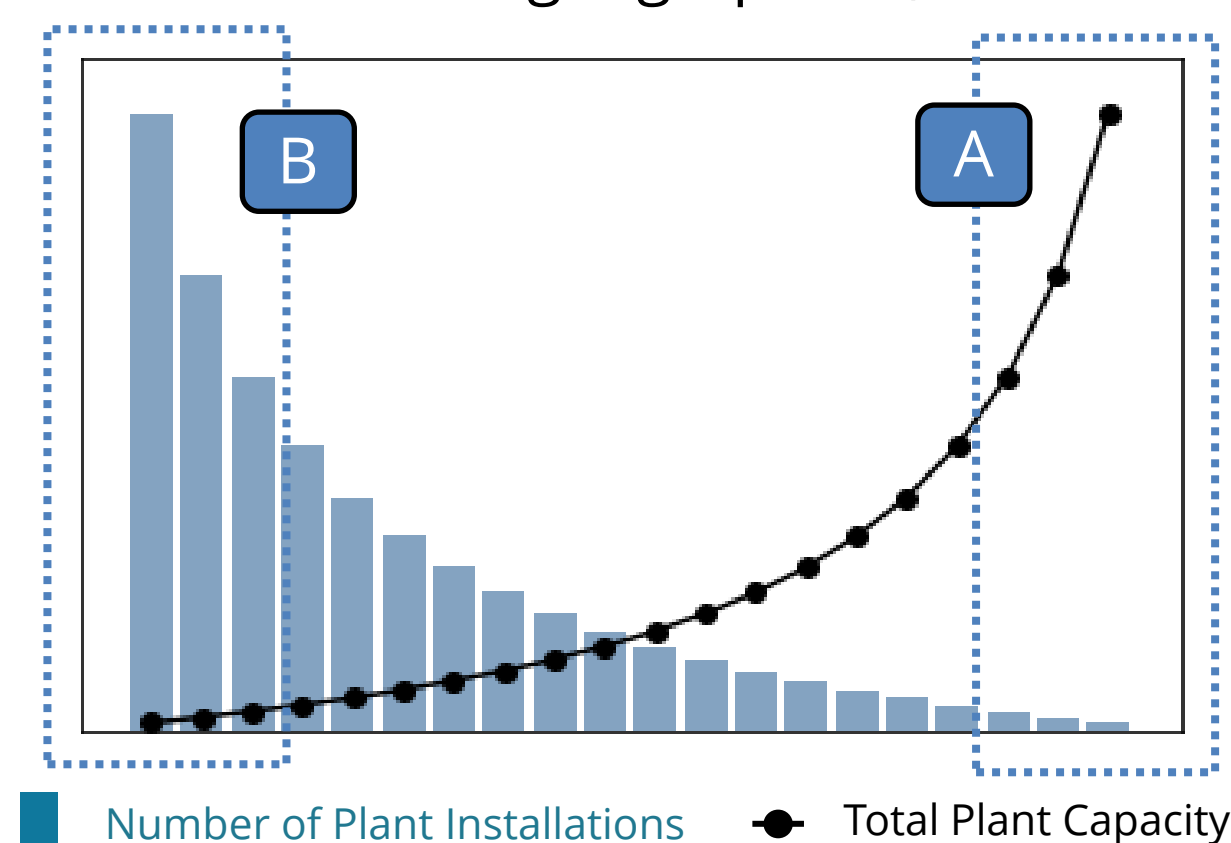
Optimal Design Approaches for Rapid, Cost-Effective Manufacturing and Deployment of Chemical Processes

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Motivation^[1]

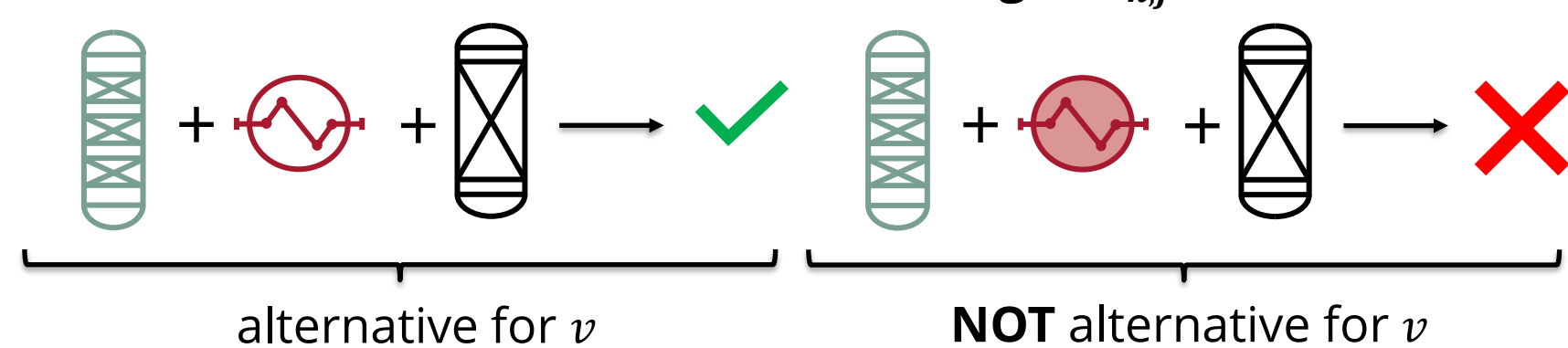
Optimally deploy a process system across decentralized sites with different geographical, environmental & operating requirements



Process Family Design includes the benefits of both

Formulation 1: Discretization Approach^[4]

V : set of process variants identified by unique performance targets & feed conditions
 K : Set of unit types considered for shared design for all variants in the process family
 A_v : Set of feasible alternatives (i.e. combination of designs $d_{k,j}$) for a variant $v \in V$



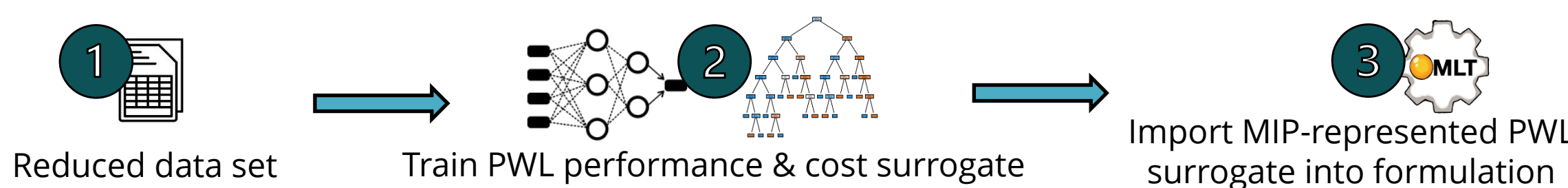
Minimize the total weighted cost of all variants in the process family

$$\begin{aligned} \min. & \sum_{v \in V} w_v \sum_{a \in A_v} c_{v,a} x_{v,a} \\ \text{s.t.} & \sum_{j \in J_k} z_{k,j} \leq N_k \quad \forall k \in K \quad \text{Select units for manufacture} \\ & \sum_{a \in A_v} x_{v,a} = 1 \quad \forall v \in V \quad \text{Select 1 alternative} \\ & x_{v,a} \leq z_{k,j} \quad \forall v \in V, a \in A_v, (k,j) \in Q_a \quad \text{Alternative must be manufactured} \\ & z_{k,j} \in \{0,1\} \quad \forall k \in K, j \in S_j \\ & 0 \leq x_{v,a} \leq 1 \quad \forall v \in V, a \in A_v \end{aligned}$$

At optimality, the solution will converge to binary under mild assumptions

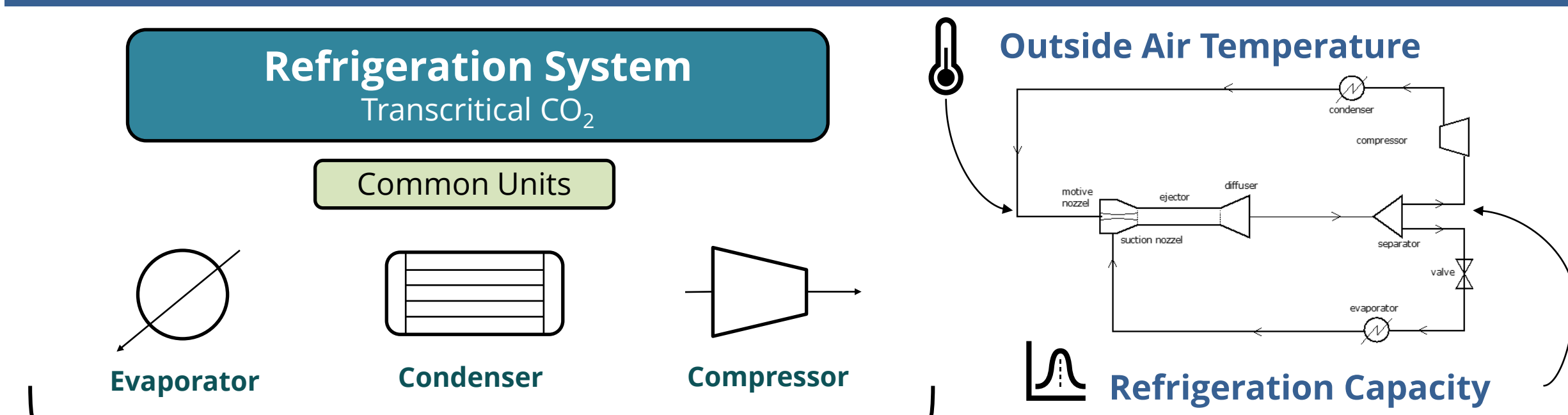
Formulation 2: Piecewise Linear Surrogates^[5]

Formulation 1	Formulation 2
<ul style="list-style-type: none"> Computational expensive simulations. Discrete sets limit optimal design decisions. 	<ul style="list-style-type: none"> Order of magnitude less data required. Approx. continuous design space.

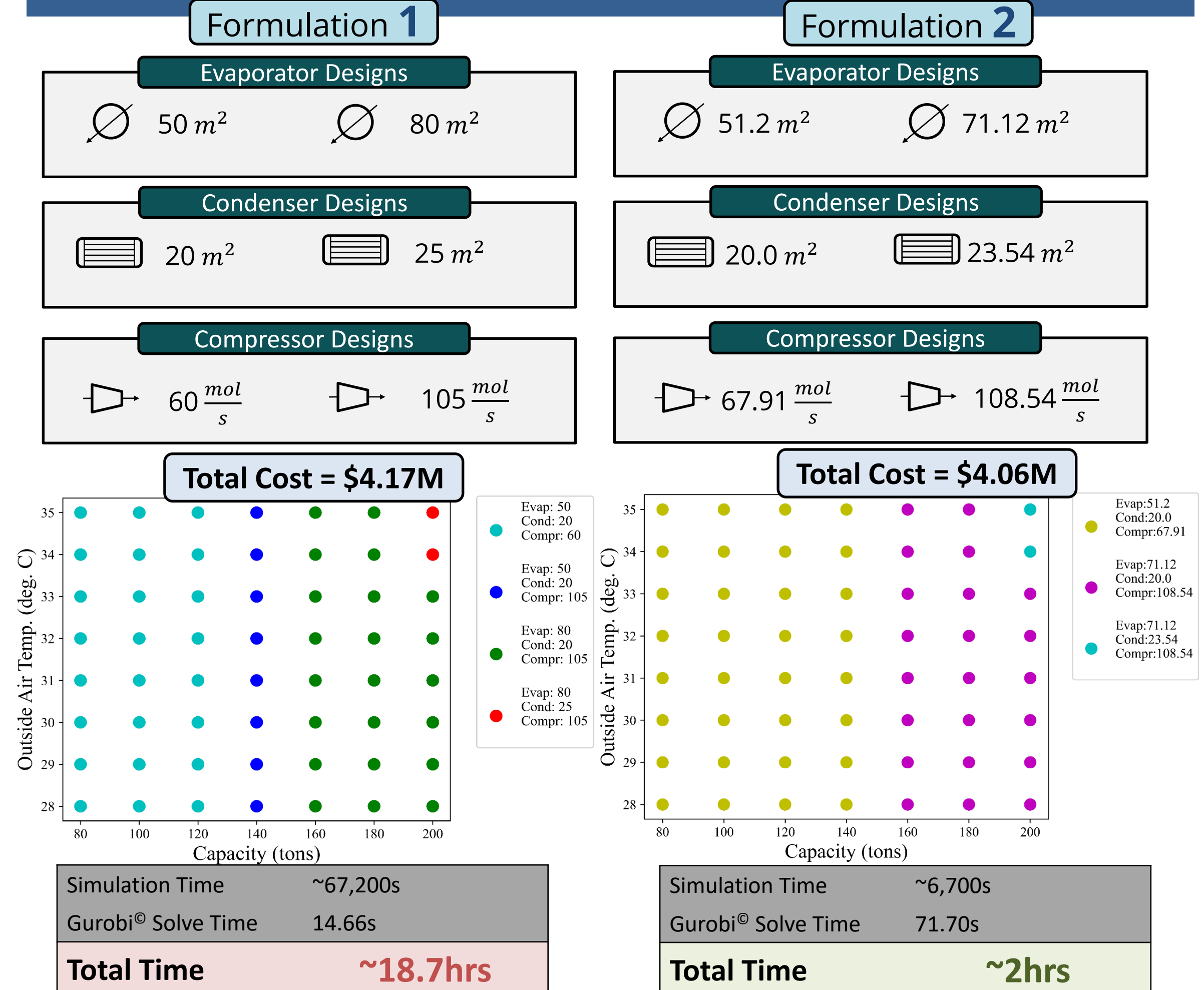


$$\begin{aligned} \min. & \sum_{i \in I} w_i c_i \\ \text{s.t.} & c_i = g^c(b_i, d_{i,1}, \dots, d_{i,k}) \quad \forall i \in I \quad \text{Cost surrogate} \\ & p_i = g^p(b_i, d_{i,1}, \dots, d_{i,k}) \quad \forall i \in I \quad \text{Performance surrogate} \\ & \bigvee_{j \in J_k} \begin{bmatrix} Y_{i,k,j} \\ d_{i,k} = \hat{d}_{k,j} \end{bmatrix} \quad \forall i \in I, k \in K \quad \text{Assignment of unit designs to variant} \\ & \hat{d}_{k,j}^{LB} \leq \hat{d}_{k,j} \leq \hat{d}_{k,j}^{UB} \quad \forall k \in K, j \in J_k \quad \text{Design boundaries} \\ & p_i^{LB} \leq p_i \leq p_i^{UB} \quad \forall i \in I \quad \text{Performance boundaries} \\ & \hat{d}_{k,j-1} \leq \hat{d}_{k,j} \quad \forall k \in K, j \in J_k: j > 1 \\ & Y_{i,k,j} \in \{\text{True}, \text{False}\} \quad \forall i \in I, k \in K, j \in J_k \end{aligned}$$

Case Study^[6]

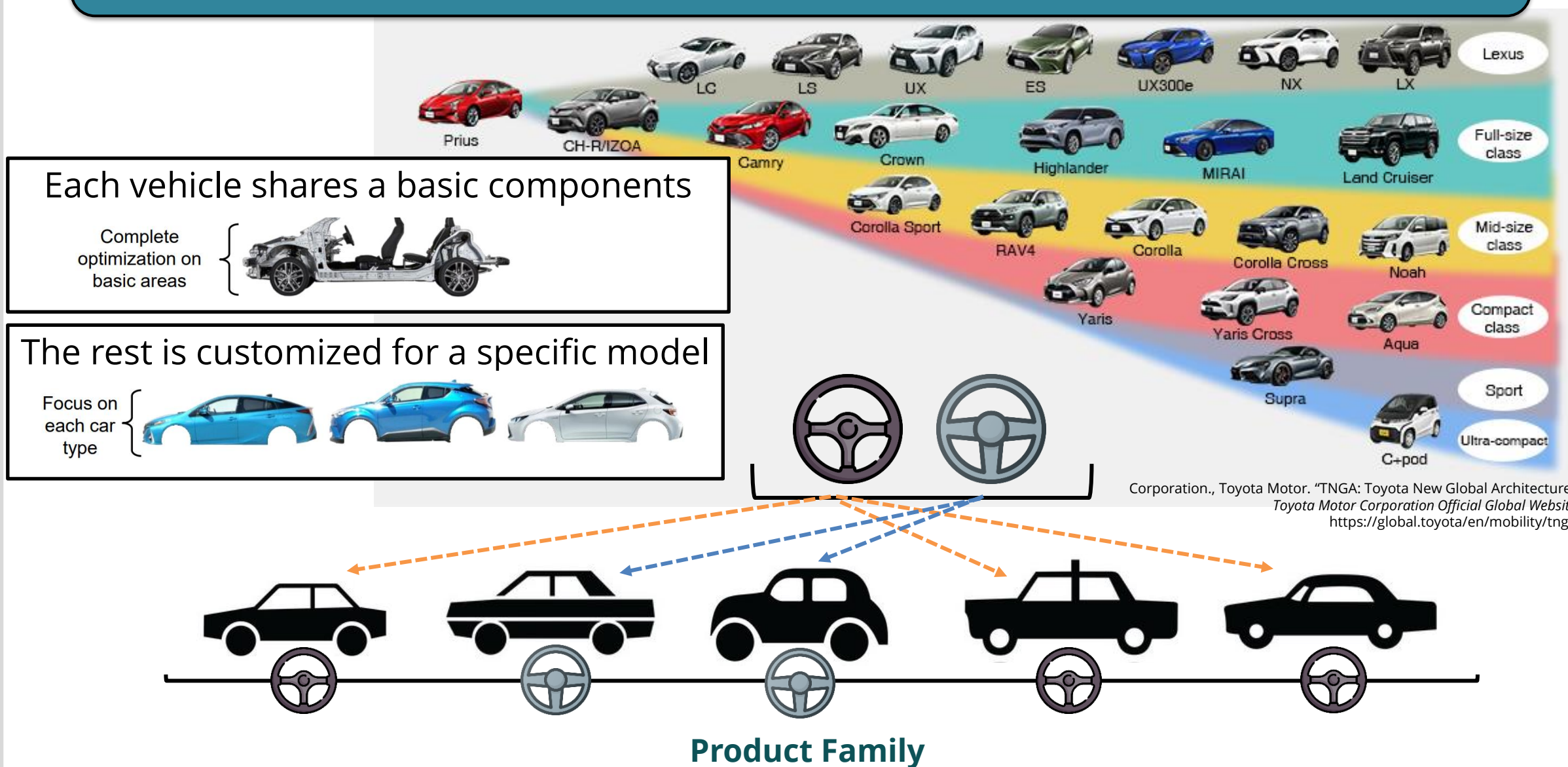


Results^[5]

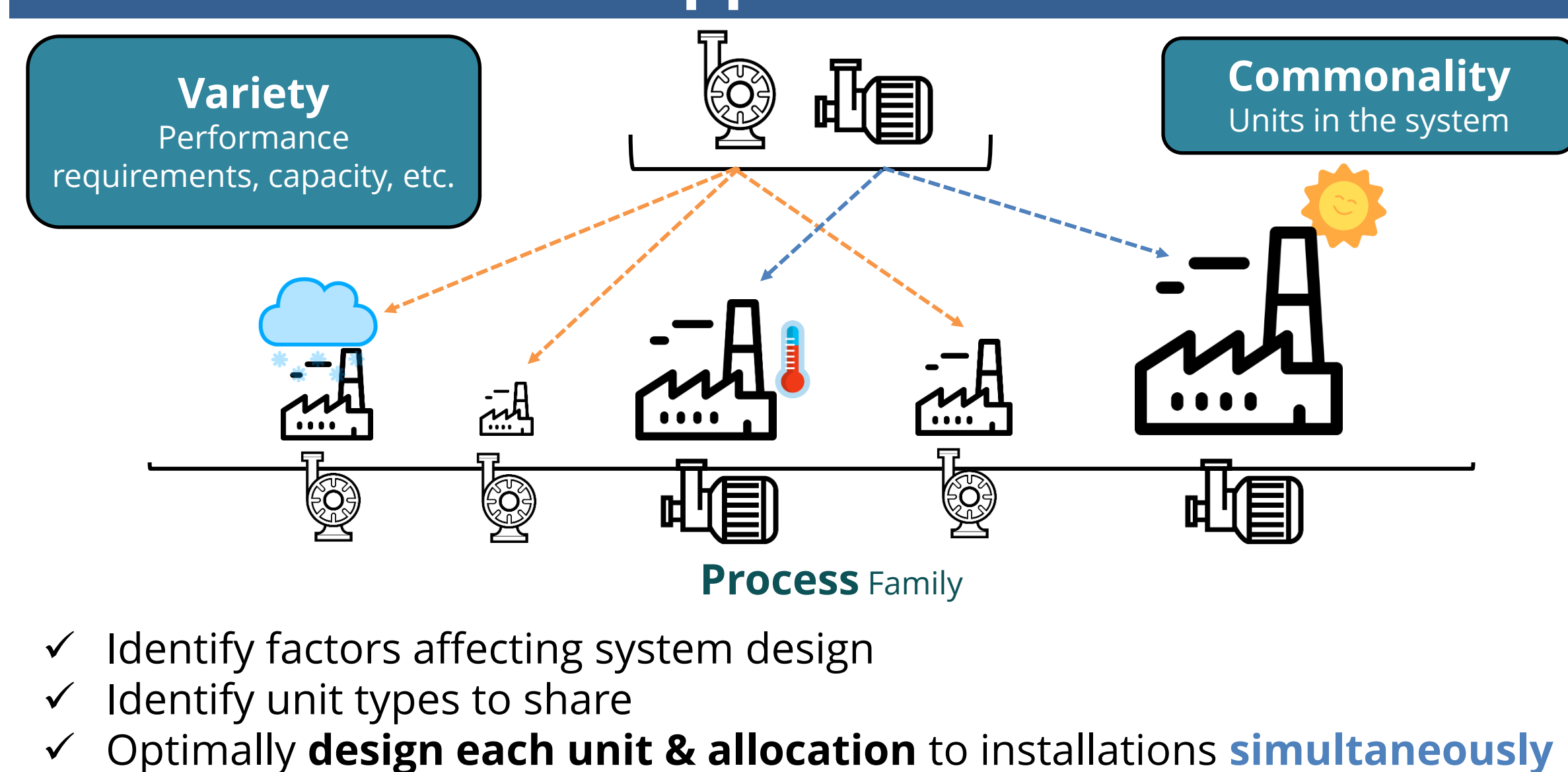


Background: Product Family Design^[2,3]

A set of products that share one or more common "element(s)" yet target a variety of different market segments



Approach



Conclusions

- ✓ Reduced manufacturing costs
 - Economies of numbers (modular concepts at unit level)
 - Economies of scale (customization to design range)
- ✓ Reduced engineering design costs
- ✓ Multiple scalable optimization formulations^[4,5]
- ✓ **Order of magnitude reduction** in data requirements^[5]
- ✓ Economies of numbers yields **approx. 8% savings** of projected **total annualized cost**^[7,8]

Future Work

- Incorporate economies of numbers savings into formulation
- Expand to more large-scale industrial case studies (decomposition)

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