

Über die Dimensionierung von Energiebereitstellungssystemen in produzierenden Unternehmen

(On the dimensioning of energy conversion systems for manufacturing companies)

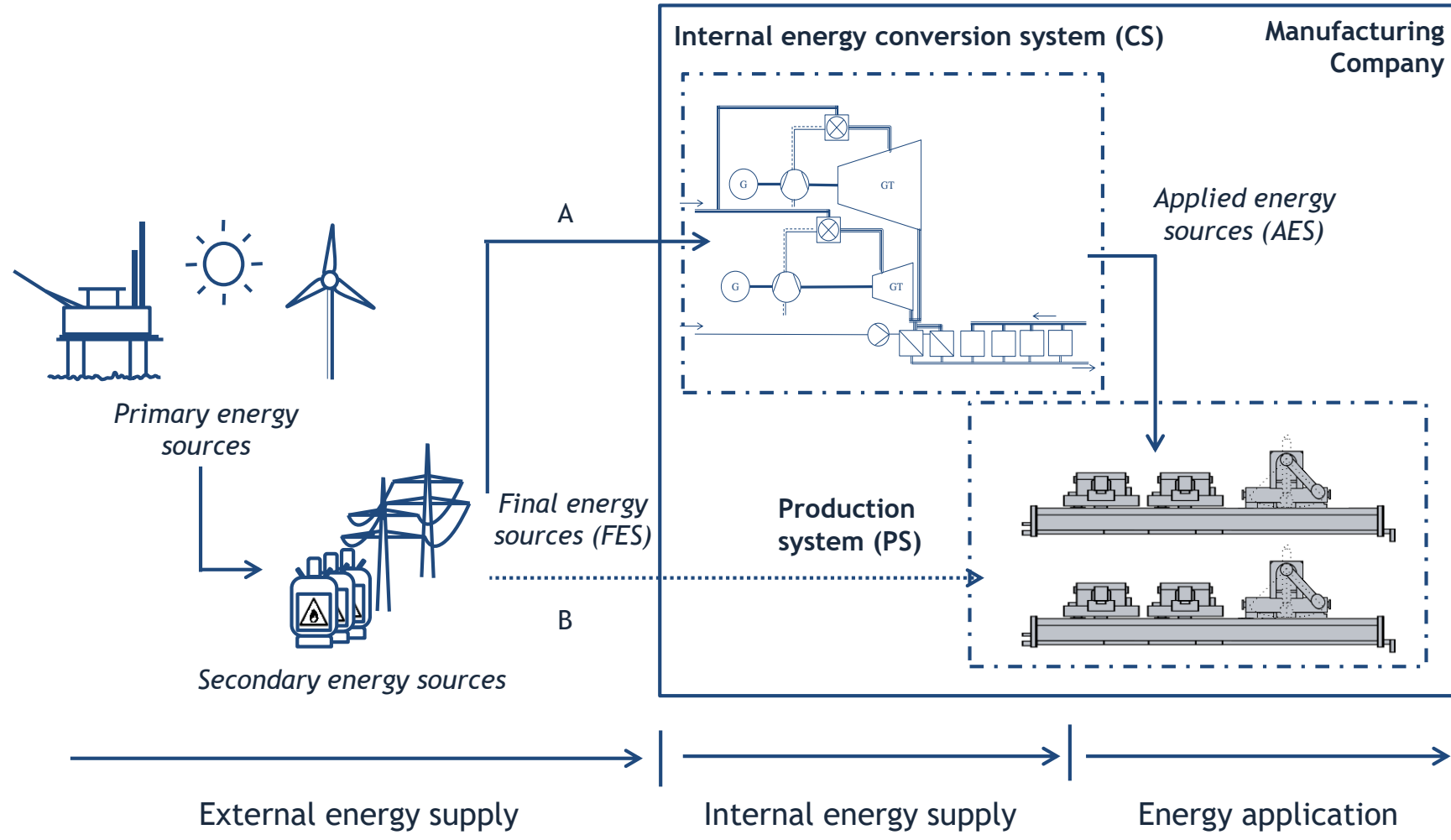
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1. Motivation



(Gahm et al. 2016)

1. Motivation
2. Decision problem with special aspects
3. Planning approach and implementation
4. Results
5. Further research
6. References

2. Decision problem with special aspects

Focus

- (internal) Conversion system transforming FES to AES

Special aspects

- Interdependencies between the CS and PS
- Technical characteristics of conversion units (CUs), particularly (part load) conversion efficiencies

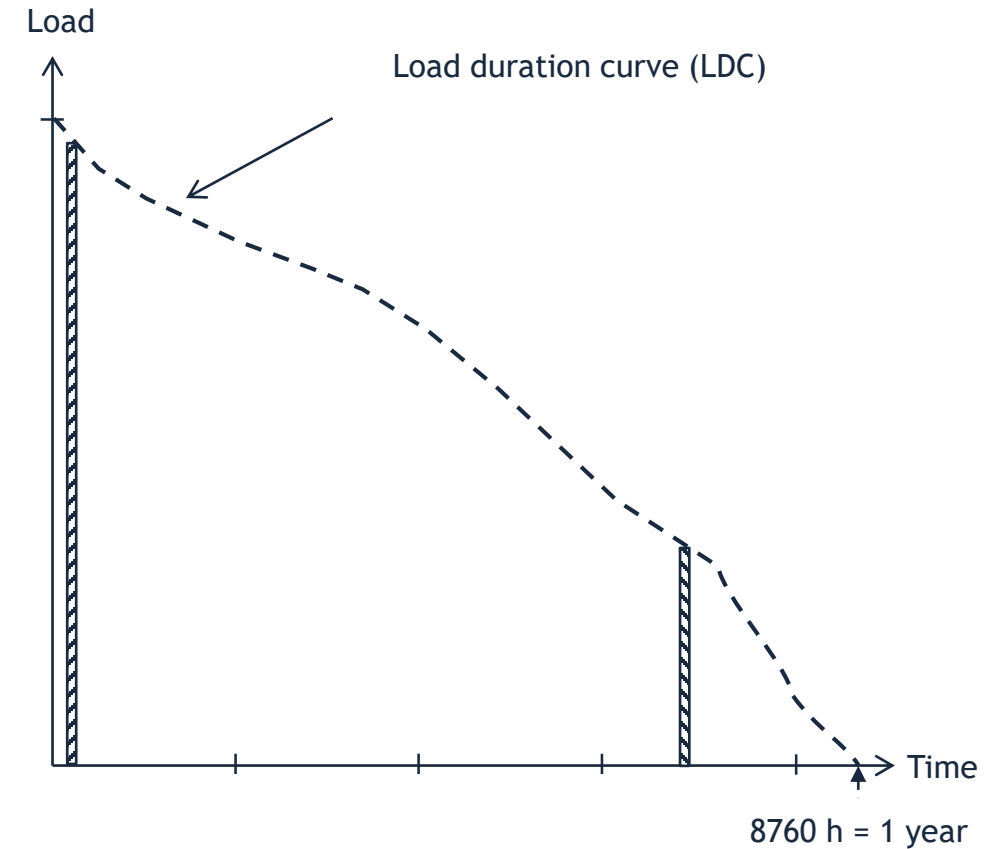
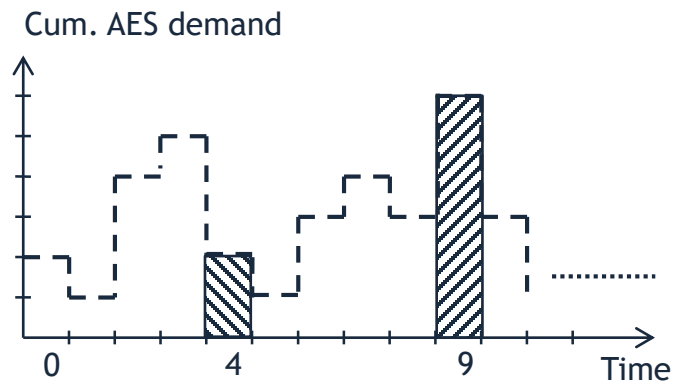
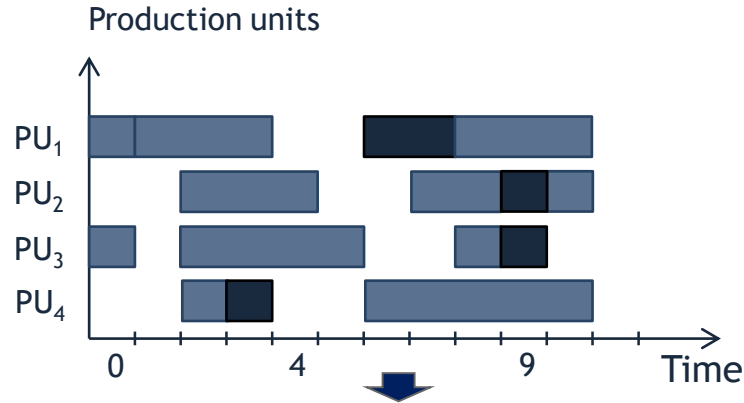
Decision problem:

The simultaneous dimensioning and configuration of energy conversion units for manufacturing companies to improve long-term energy efficiency.

2.1 Load duration curves (LDC)

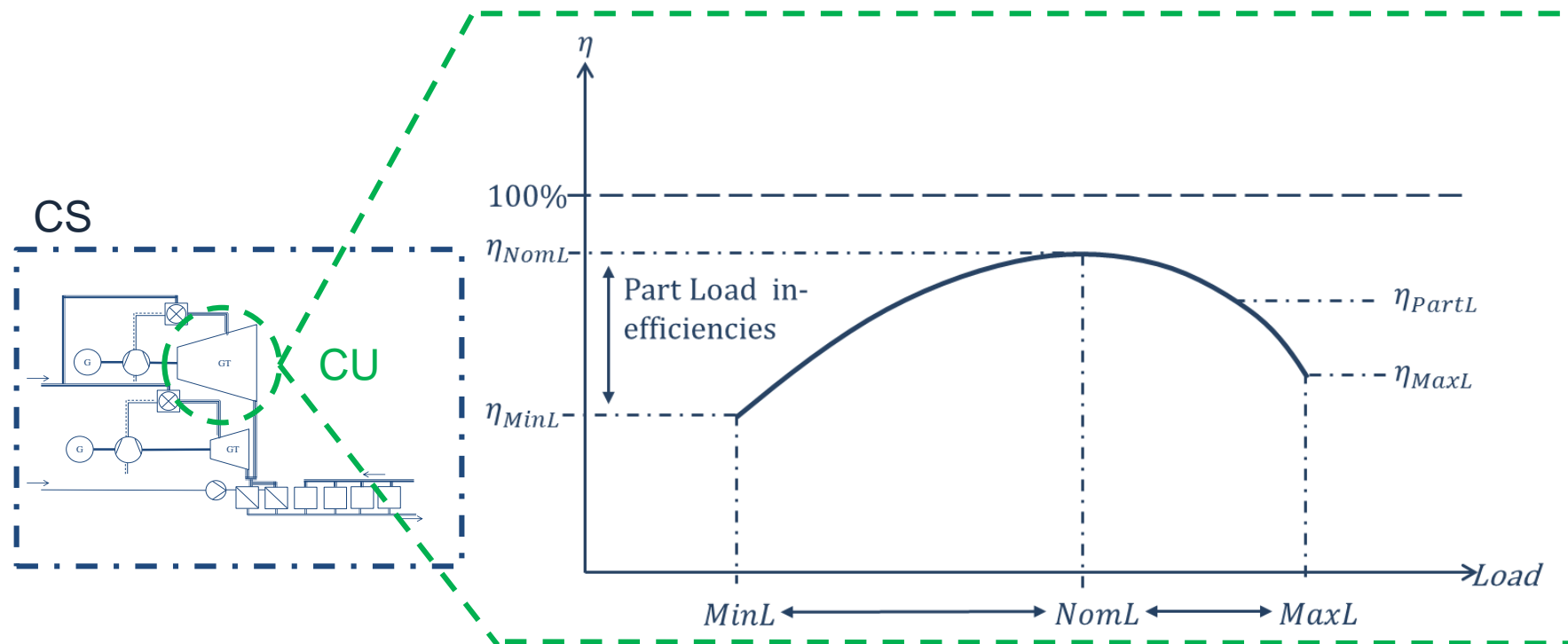
■ PS: Operational scheduling

→ Load duration curves (LDC) represent the (historical) AES demand

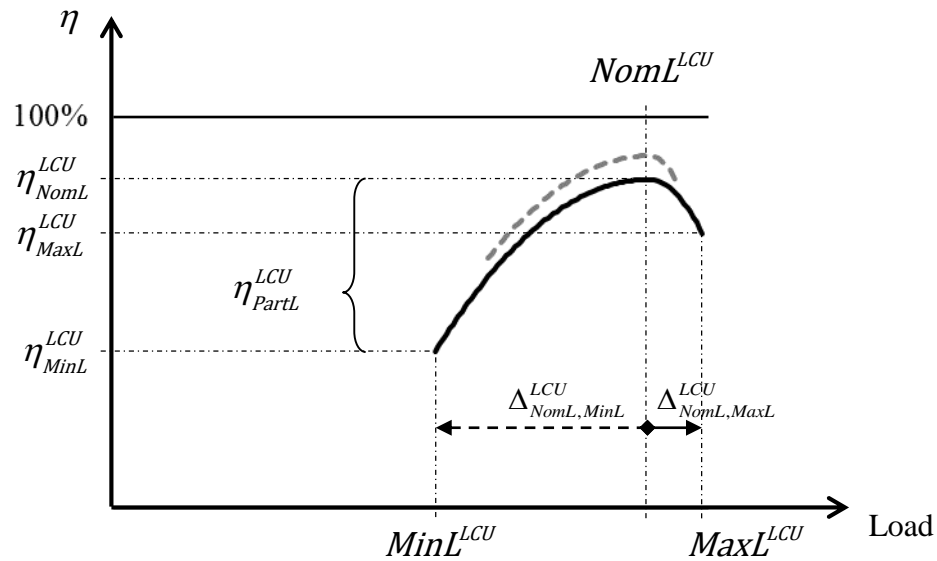


A CS consists of one or more conversion units (CU) with individual characteristics:

- Minimal, nominal and maximal load ($MinL$, $NomL$, $MaxL$)
- Load efficiencies (η_{MinL} , η_{NomL} , η_{MaxL})

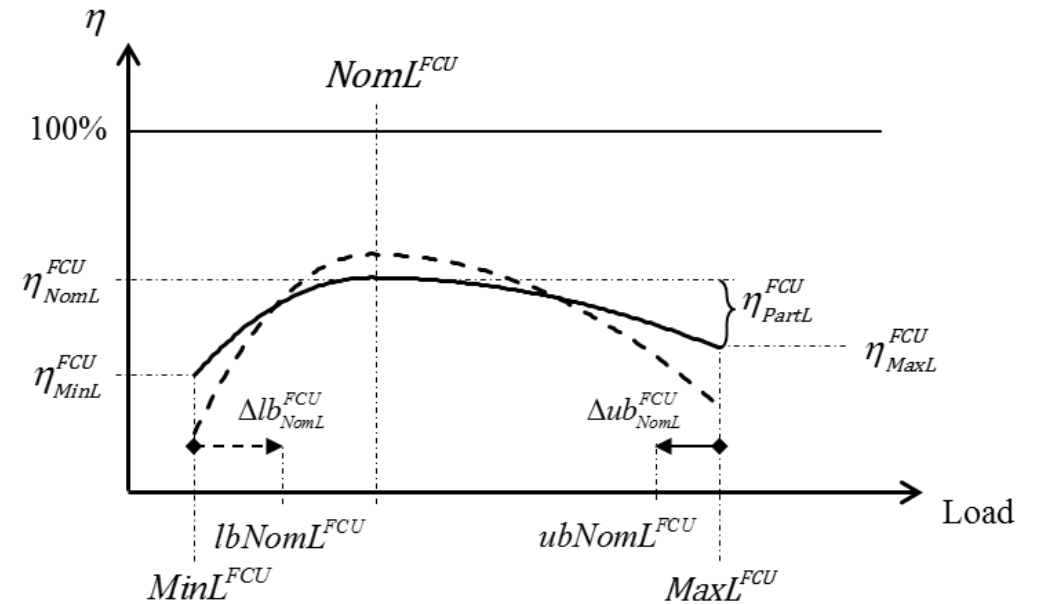


2.3 Basic types of CUs



Large conversion unit (LCU)

- Fulfills basic AESD with max. efficiency
→ high $NomL$ efficiency, small control range
- Based on $NomL$ $MaxL$ and $MinL$ are derived and thus, the dimensions the CU



Flexible conversion unit (FCU)

- Handles AESD peaks
→ smaller $NomL$ efficiency than LCU but a wider control range
- The $NomL$ of FCUs is configured within a given range based on its $MaxL$ and $MinL$

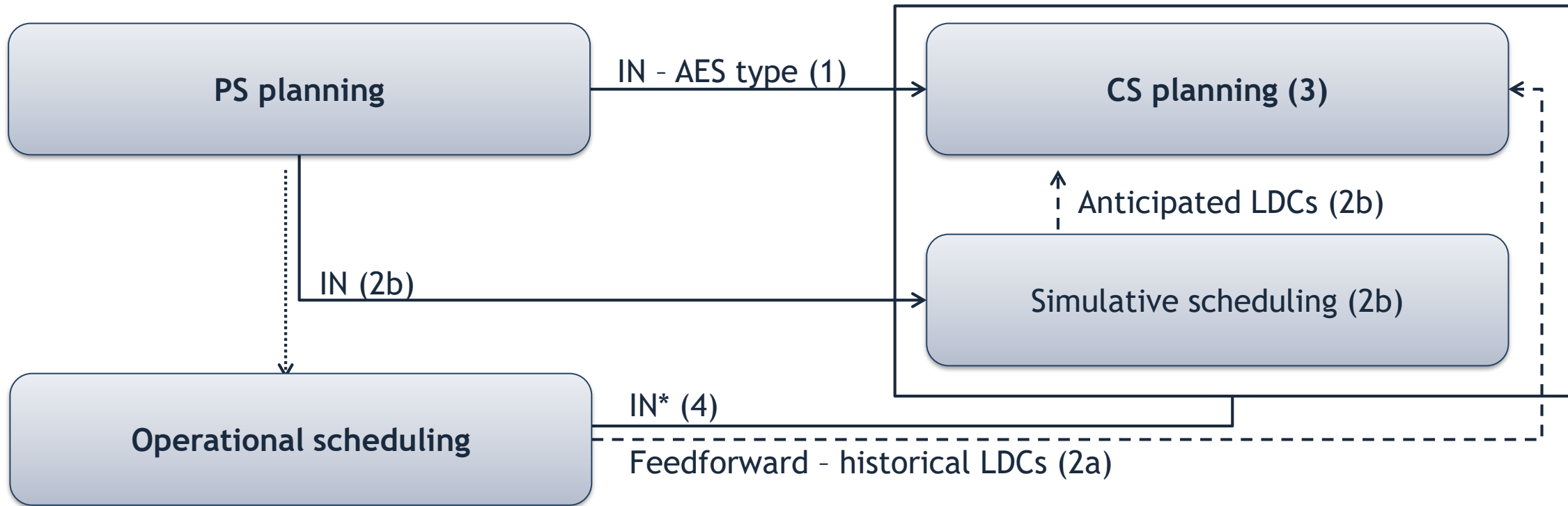
3. Planning approach and implementation

3.1 Decision environment

3.2 Simulative scheduling (2b) - LDC generation

3.3 CS planning (3) - solution methods

3.1 Decision environment



(based on „ Distributed decision making—a unified approach” by Schneeweiss 2003)

→ Simulative Scheduling is used to anticipate LDCs if no historical data is available or new scheduling objectives are in mind

3.2 Simulative scheduling (2b) - LDC generation

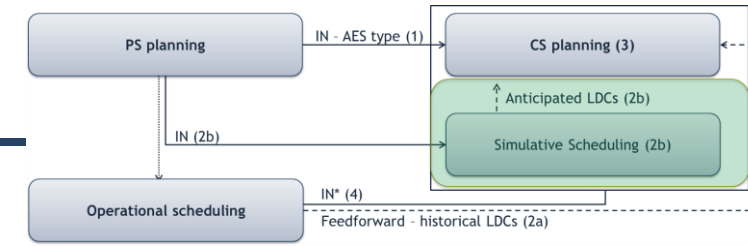
- Simulative Scheduling (2b) to anticipate LDCs

- 240 schedules per year with ~480 minutes/day

(For feasibility the planning horizon T is

$$T = \max\{480, \text{Makespan}^{LPT}\}$$

- Different production environments
 - Different energy demand characteristics
 - Scheduling with three different objectives



- Company size m (S/M)
- Product complexity p (MS / FC)
- Number of jobs n

Size	p	m	n	
S	MS	30	3	[44, 48]
			4	[58, 64]
	FC	80	3	[14, 18]
			4	[18, 24]
M	MS	30	10	[145, 160]
			12	[174, 192]
	FC	80	10	[45, 60]
			12	[54, 72]

→ 4 production environments

3.2 Simulative scheduling (2b) - LDC generation

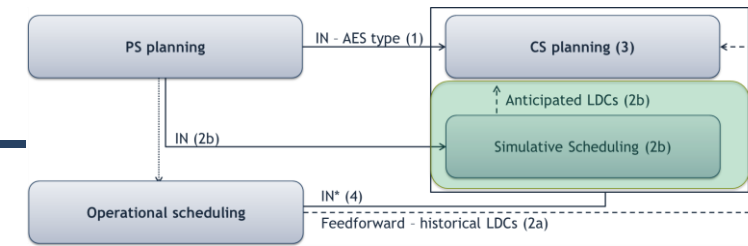
- Simulative Scheduling (2b) to anticipate LDCs

- 240 schedules per year with ~480 minutes/day

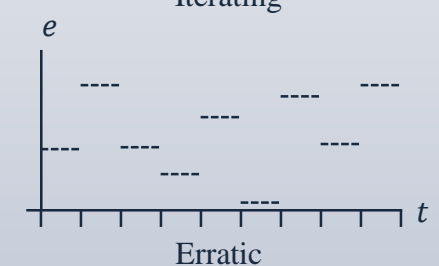
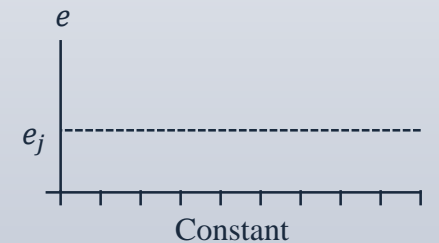
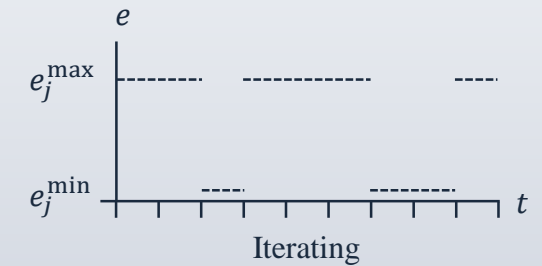
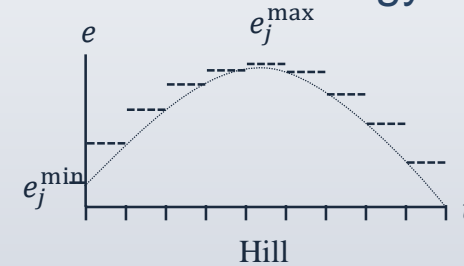
(For feasibility the planning horizon T is

$$T = \max\{480, \text{Makespan}^{LPT}\}$$

- Different production environments
- Different energy demand characteristics
- Scheduling with three different objectives



- Four divers energy demand curves



- Two energy ranges (e_j^{\min} to e_j^{\max})

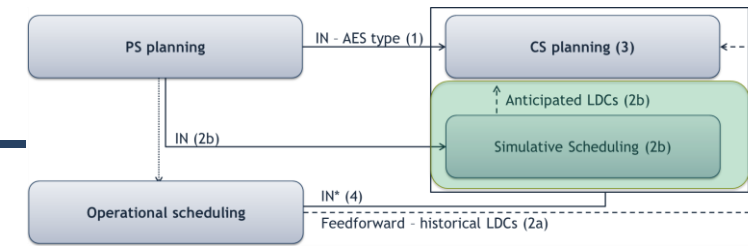
- Small range (SR)
 - Long range (LR)

→ 8 energy settings

3.2 Simulative scheduling (2b) - LDC generation

- Simulative Scheduling (2b) to anticipate LDCs
 - 240 schedules per year with ~480 minutes/day
(For feasibility the planning horizon T is
 $T = \max\{480, \text{Makespan}^{LPT}\}$)
 - Different production environments
 - Different energy demand characteristics
 - Scheduling with three different objectives

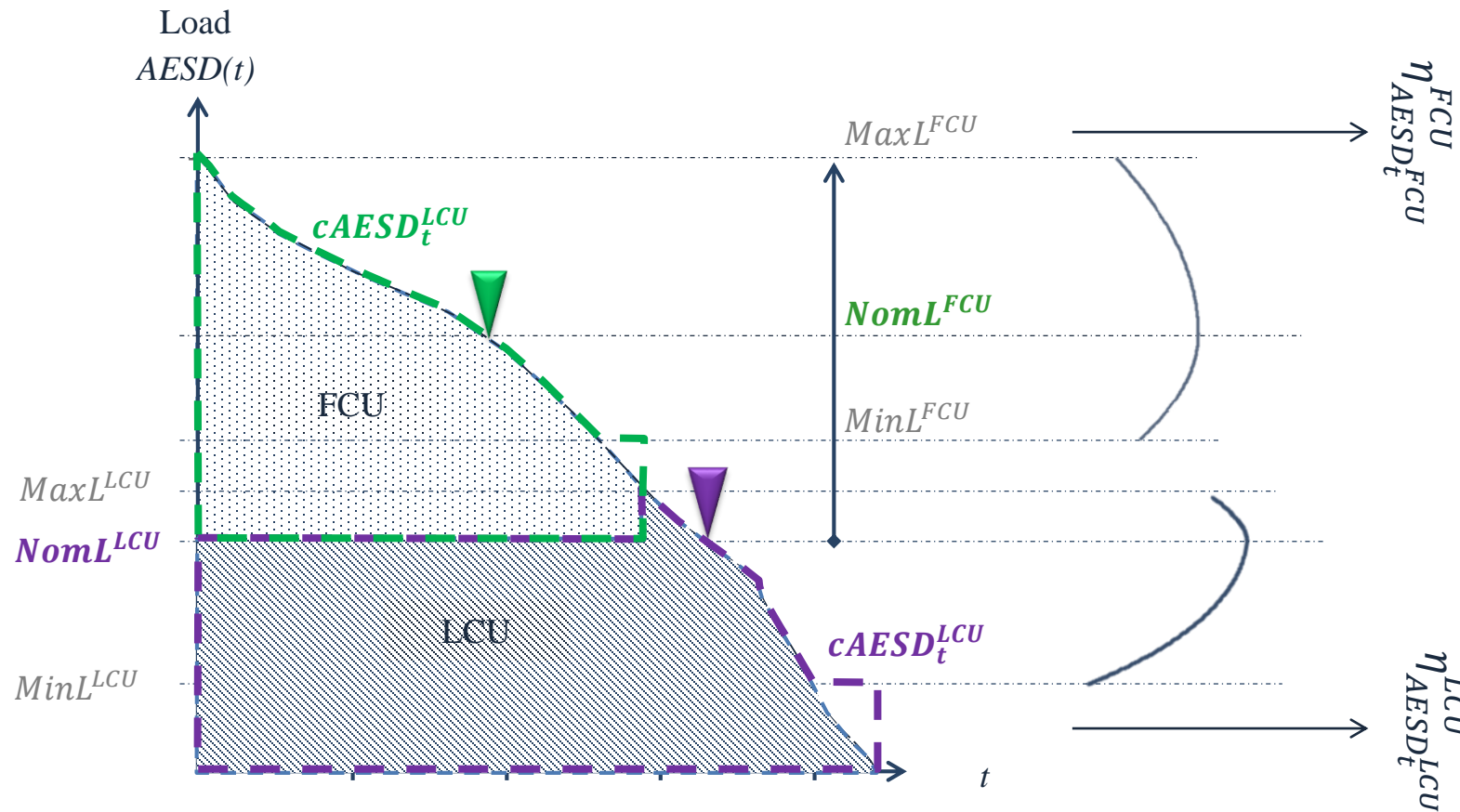
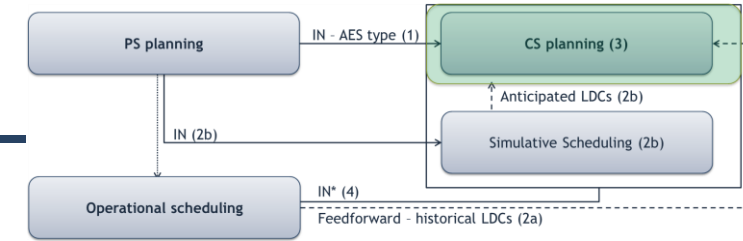
- Makespan (C_{max})
- Total flow time (TFT)
- AESD leveling ($SQmean = \text{Min} \sum (e_t - \bar{e})^2$)



- ➔ 4 production environments combined with 8 energy settings map 32 company types
- ➔ The 3 objectives lead to 96 anticipated LDC (3 per company type)

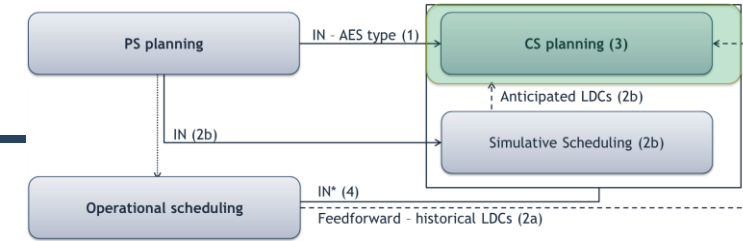
3.3 CS planning (3) - solution methods

- Assuming an CS with two CUs (one LCU and one FCU):



$$\text{Min} \sum_{t=1}^T \frac{cAESD_t^{LCU}}{\eta_{AESD_t^{LCU}}^{LCU}} + \frac{cAESD_t^{FCU}}{\eta_{AESD_t^{FCU}}^{FCU}}$$

3.3 CS planning (3) - solution methods



- TEH - two-step truncated enumeration heuristic
 - First step: enumerate $NomL^{LCU}$ by relatively fixed $NomL^{FCU}$
 - Second Step: Enumerate $NomL^{FCU}$ with fixed $NomL^{LCU}$

- MINLP - mixed integer nonlinear program
 - Objective function: Minimize total FESD

$$\text{Min} \sum_{L=l}^L \left(\frac{cAESD_l^{LCU}}{\mu_l^{LCU}} + X_l * \frac{cAESD_l^{FCU}}{\mu_l^{FCU}} \right) * nop_l \quad \text{with} \quad \begin{array}{l} nop_l \text{ Number of periods with AESD level } l \\ X_l = 1 \text{ if FCU is needed and} \\ X_l = 0 \text{ if FCU is not needed to cover the } AE\text{SD}_l \end{array}$$

- Determination of the part load efficiencies is not linear

$$\begin{aligned} \mu_l^{LCU} \leq Y_l^{LCU} & * \left(\frac{\mu_{MaxL}^{LCU} - \mu_{NomL}^{LCU}}{(MaxL^{LCU} - NomL^{LCU})^2} \right) * (cAESD_l^{LCU} - NomL^{LCU})^2 + \mu_{NomL}^{LCU} \\ & + (1 - Y_l^{LCU}) * \left(\frac{\mu_{MinL}^{LCU} - \mu_{NomL}^{LCU}}{(MinL^{LCU} - NomL^{LCU})^2} \right) * (cAESD_l^{LCU} - NomL^{LCU})^2 + \mu_{NomL}^{LCU} \end{aligned}$$

4. Results

4.1 LCU and FCU parameter-setting analysis

4.2 Overall results

4.1 LCU and FCU parameter-setting analysis

	LCU-0	LCU-1	LCU-2	LCU-3	LCU-4	LCU-5
η_{MaxL}^{LCU}	87%	87%	85%	85%	80%	91%
η_{NomL}^{LCU}	95%	95%	95%	93%	97%	93%
η_{MinL}^{LCU}	82%	82%	80%	80%	75%	86%
$\Delta_{NomL,MaxL}^{LCU}$	0.05	0.10	0.10	0.10	0.05	0.05
$\Delta_{NomL,MinL}^{LCU}$	0.30	0.40	0.40	0.40	0.30	0.30

	FCU-0	FCU-1	FCU-2	FCU-3	FCU-4
η_{MaxL}^{FCU}	65%	65%	65%	60%	70%
η_{NomL}^{FCU}	84%	84%	82%	86%	82%
η_{MinL}^{FCU}	60%	60%	60%	55%	65%
$\Delta_{MaxL, NomL}^{FCU}$	0,15	0,05	0,05	0,15	0,15
$\Delta_{MaxL,MinL}^{FCU}$	0,30	0,10	0,10	0,30	0,30

Large conversion unit (LCU)

- one LCU as basis of comparison (LCU-0)
- increased operational range (LCU-1)
- increased operational range & modified efficiencies (LCU-2, LCU-3)
- modified efficiencies (LCU-4, LCU-5)
- → 6 divers LCUs

Flexible conversion unit (FCU)

- one FCU as basis of comparison (FCU-0)
- looser bounds of *NomL* (FCU-1)
- looser bounds of *NomL* & adjusted *NomL* efficiency (FCU-2)
- modified efficiencies (FCU-3, FCU-4)
- → 5 divers FCUs

4.1 LCU and FCU parameter-setting analysis

FCU parameters analysis

- FCU-3 is most preferable for almost all company types (28 of 32)
(FCU-3: increased NomL efficiency, but suffers in part-load efficiency)
- FCU-1 (3 of 32) and FCU-4 (1 of 32) are more suitable for specific company types
(FCU-1: looser bounds for the NomL)
(FCU-4: increased part-load efficiency, but suffers in NomL efficiency)

	FCU-0	FCU-1	FCU-2	FCU-3	FCU-4
η_{MaxL}^{FCU}	65%	65%	65%	60%	70%
η_{NomL}^{FCU}	84%	84%	82%	86%	82%
η_{MinL}^{FCU}	60%	60%	60%	55%	65%
$\Delta_{MaxL, NomL}^{FCU}$	0,15	0,05	0,05	0,15	0,15
$\Delta_{MaxL, MinL}^{FCU}$	0,30	0,10	0,10	0,30	0,30

LCU parameters analysis

- LCU-5 is not preferable due to its lower nominal load efficiency (0 of 32)
(LCU-5: increased part-load efficiency, but suffers in NomL efficiency)
- although LCU-4 has the highest nominal load efficiency (11 of 32) , LCU-1 with its larger operational range is preferable for most company types (21 of 32)

	LCU-0	LCU-1	LCU-2	LCU-3	LCU-4	LCU-5
η_{MaxL}^{LCU}	87%	87%	85%	85%	80%	91%
η_{NomL}^{LCU}	95%	95%	95%	93%	97%	93%
η_{MinL}^{LCU}	82%	82%	80%	80%	75%	86%
$\Delta_{NomL, MaxL}^{LCU}$	0.05	0.10	0.10	0.10	0.05	0.05
$\Delta_{NomL, MinL}^{LCU}$	0.30	0.40	0.40	0.40	0.30	0.30

➔ nominal load efficiency of a CU is not the only decisive parameter

➔ A CUs operational range is important

4.2 Overall results

- List of Most preferable parameters by company type
 - ➔ depending on a manufacturing company's characteristics, individual combinations of a scheduling objective and CU parameters are most suitable to maximize its energy efficiency
- LCU parameters have a greater influence than FCU parameters
- ANTIGONE (MINLP) & TEH solve with reasonably good solution quality
- SQM has the best mean and the most stable positive influence on energy efficiency
 - ➔ Testing of more energy-related scheduling objectives advisable

5. Further research

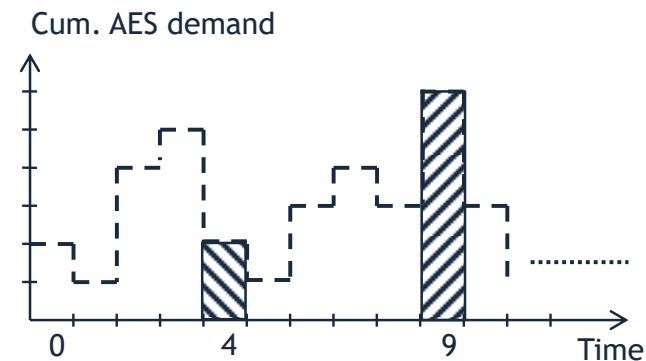
5.1 CU state and load transitions

5.2 Modeling of additional energy requirements

5.1 CU state and load transitions

- CU can be in states
 - “operating”, “off-cold”, “cold-startup”, “off-warm”, “warm-startup”, “on”
 - Predefined sequences of CU states
 - State- and/or time-dependent transitions
 - Minimum CU state durations
- While CU is in the “operating” state it delivers various AESDs (part loads)
 - Arbitrarily large load transitions within short time are not possible (Restricted ramp-ups/ ramp-downs)
 - Minimum CU part-load durations

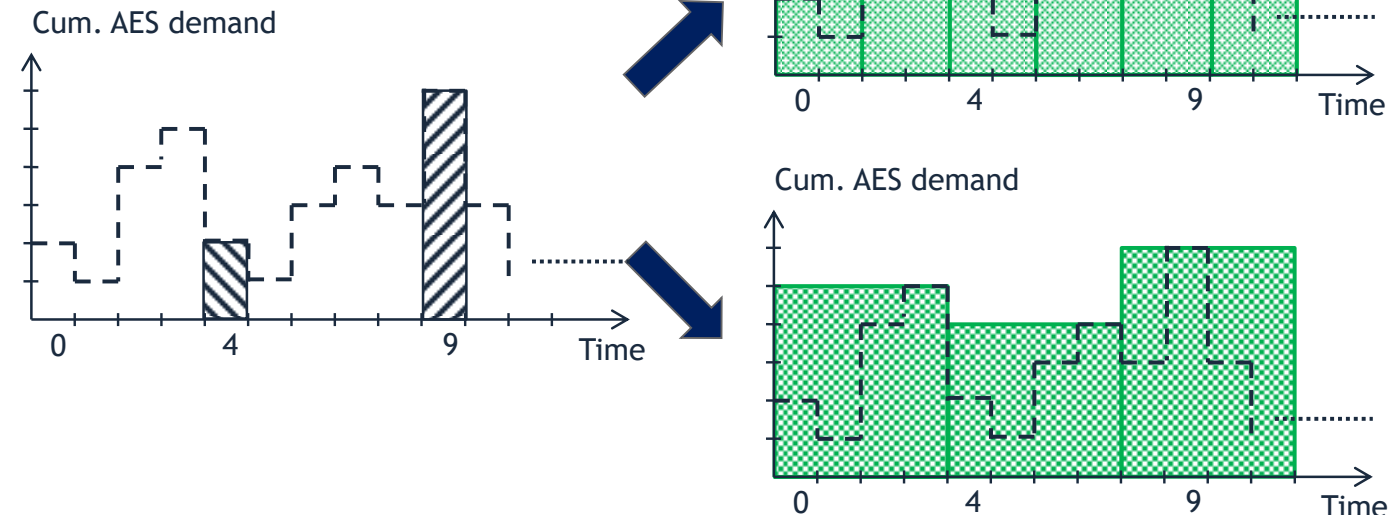
➔ both cause additional energy requirements



5.2 Modeling of additional energy requirements

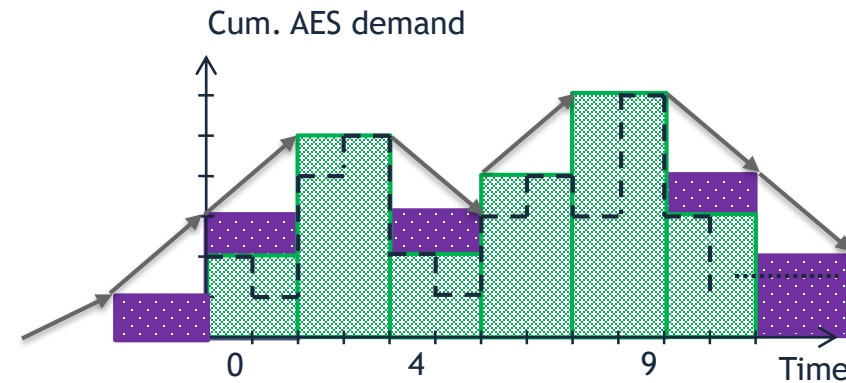
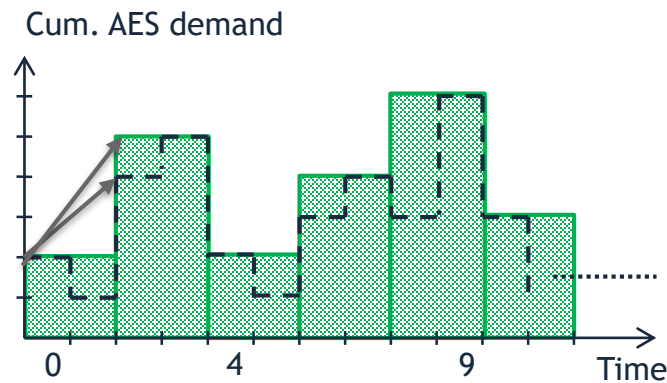
How to model additional energy requirements?

- direct consideration within the optimization model
 - level aggregation no longer possible
 - number of variables increases dramatically
- Indirect consideration by adapting the AESD to enforce a corresponding FESD
 - Minimum durations of CU states and part-loads can be considered by aggregation
 - → how many time periods have to be aggregated to a constant AESD-level?
 - → How should the level be chosen? Max. vs Mean



How to model additional energy requirements?

- Indirect consideration by adapting the AESD to enforce a corresponding FESD
 - Additional FESDs (e.g., for ramp-ups) have to be approximated
 - Possible height difference of load transition depends on available time for the transition
 - → What difference of AESD is manageable between aggregated time intervals?





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