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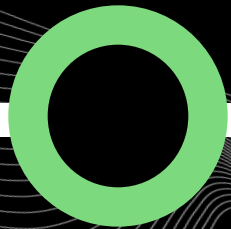
RELATING ELECTRIC VEHICLE CHARGING TO SPEED SCALING WITH JOB-SPECIFIC SPEED LIMITS

JOINT WORK WITH M.E.T. GERARDS, A. ANTONIADIS, G. HOOGSTEEEN, J. HURINK

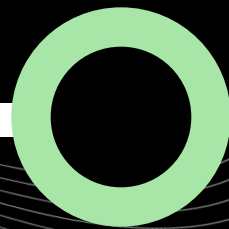
IN THIS PRESENTATION:



USE CASE



SPEED SCALING



FOCS
ALGORITHM



FOCS
PROOF SKETCH



CONCLUSION



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USE CASE - LIVING LAB



- Electric Vehicle (EV) charging infrastructure at parking lot
- ~250 charging stations
- ~450 charging stations 2030
- 900 kW peak PV
- Lease fleet

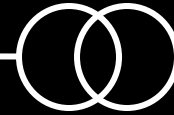


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USE CASE - LIVING LAB



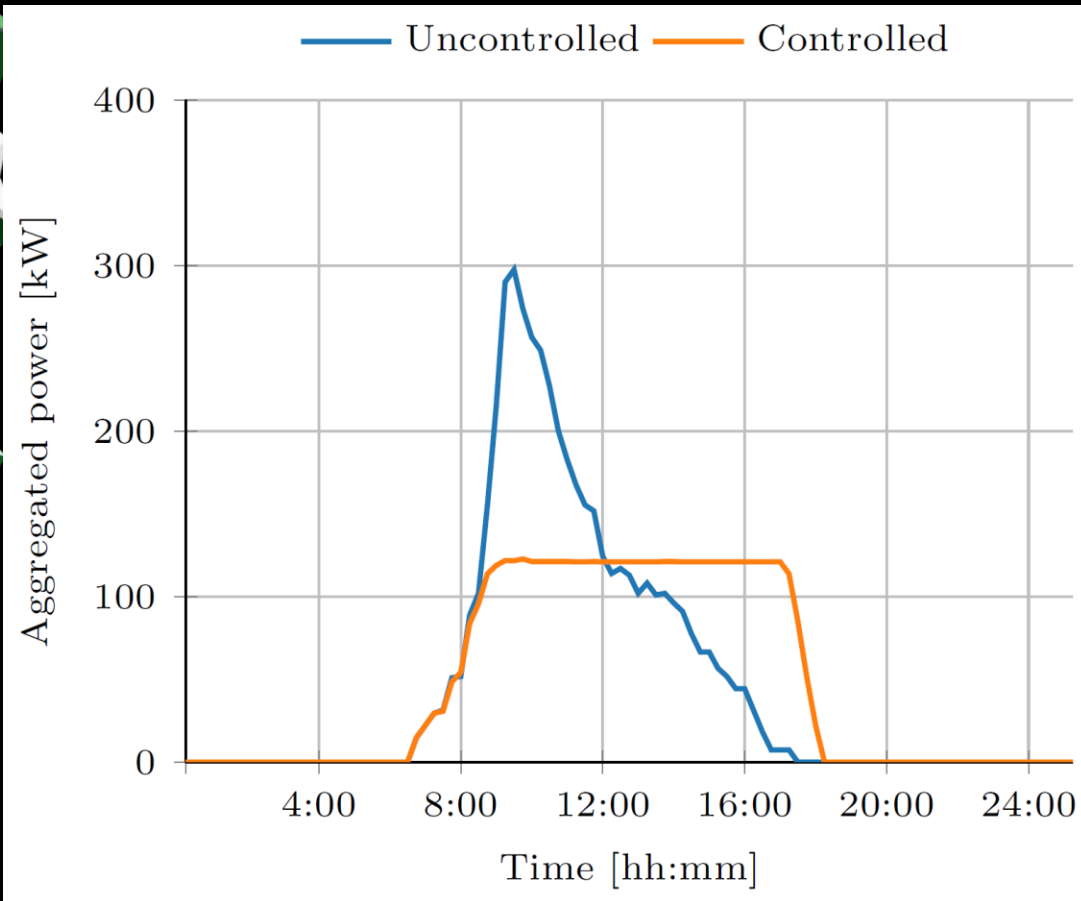
8MW (2MW)



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FOCS – L.Winschermann@utwente.nl

USE CASE - SCHEDULING



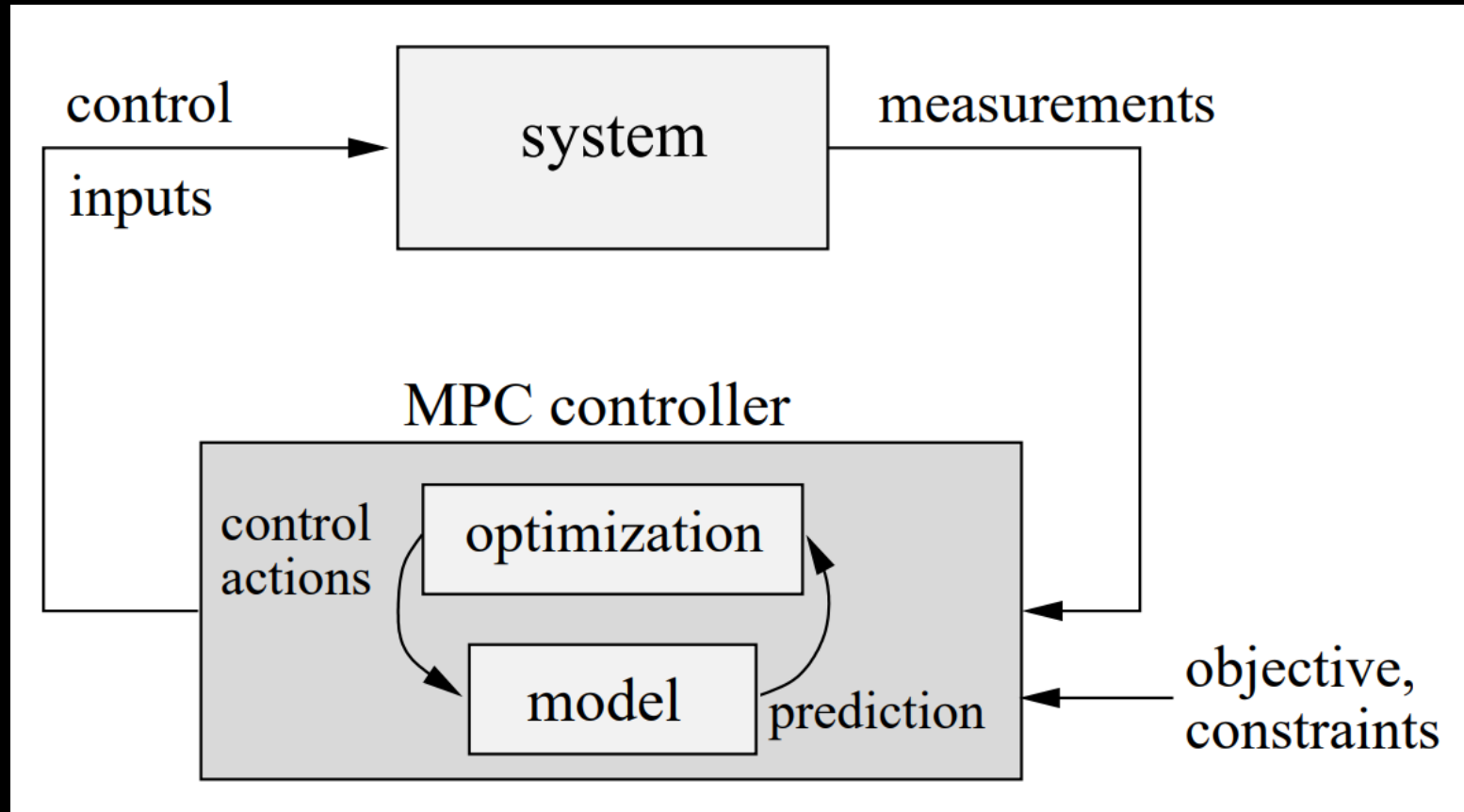
- Jobs j :
 - Arrival times
 - Departure times
 - Work (energy)
 - Speed limit (power)
- Decision variable:
 - Job schedules $s_j(t)$
- Objective function:
 - $\int_t (\sum_j s_j(t))^2 dt$

INFORMATION GAP



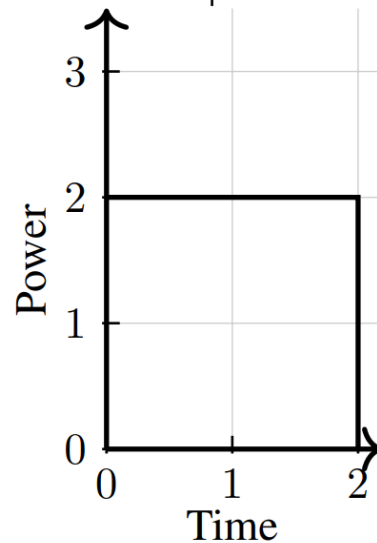
- Given in practice:
 - Arrival time (upon arrival)
 - Energy 'demand' (upon completion)
 - Departure time (upon departure)
 - Maximum power (upon query)

MODEL PREDICTIVE CONTROL

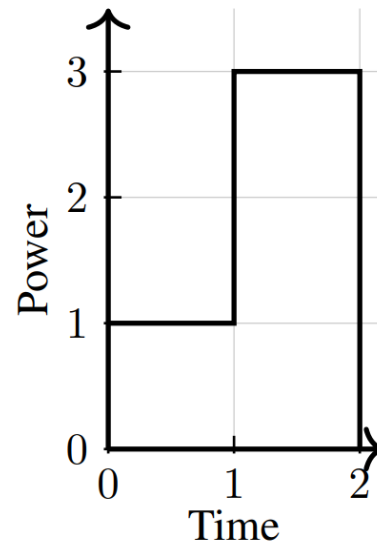


SPEED SCALING – PROCESSOR SCHEDULING

j	a_j	d_j	e_j	$p_j^{\max}$
1	0	2	2	1
2	1	2	2	2



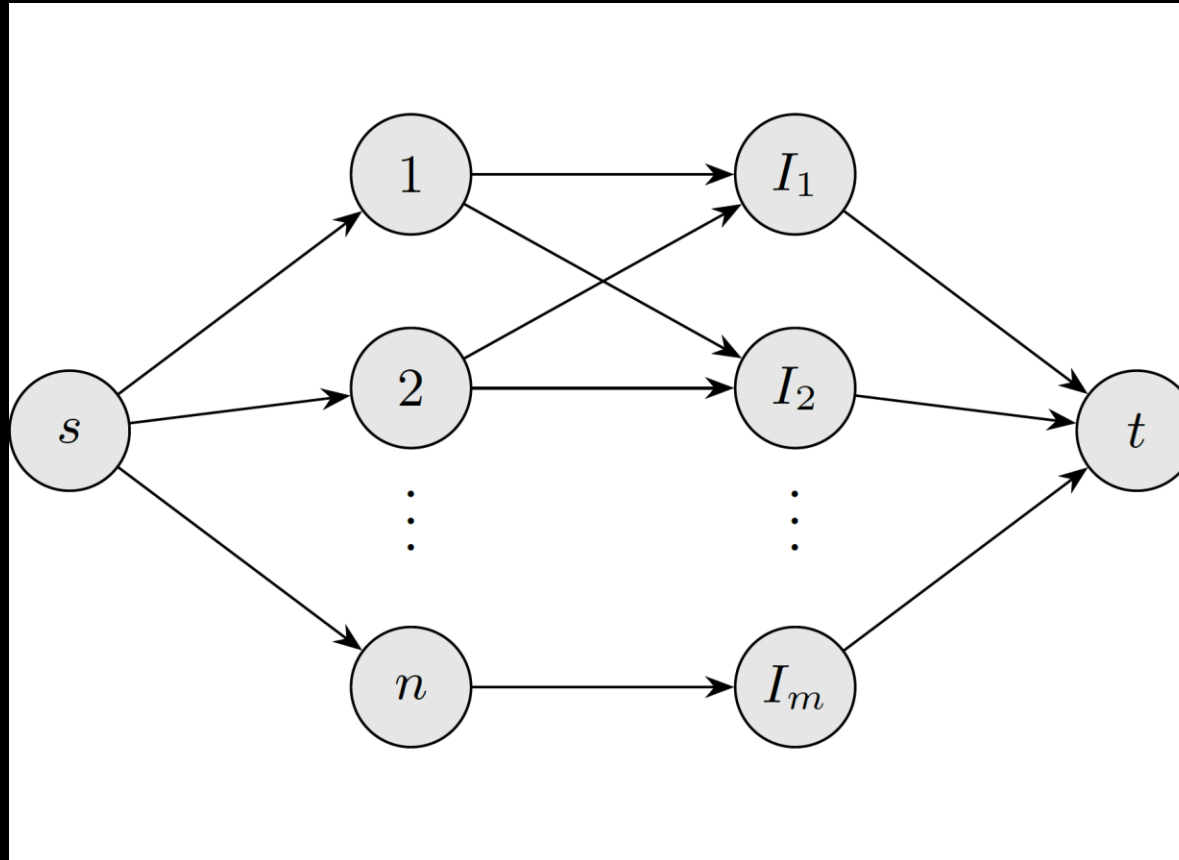
Without speed-limits



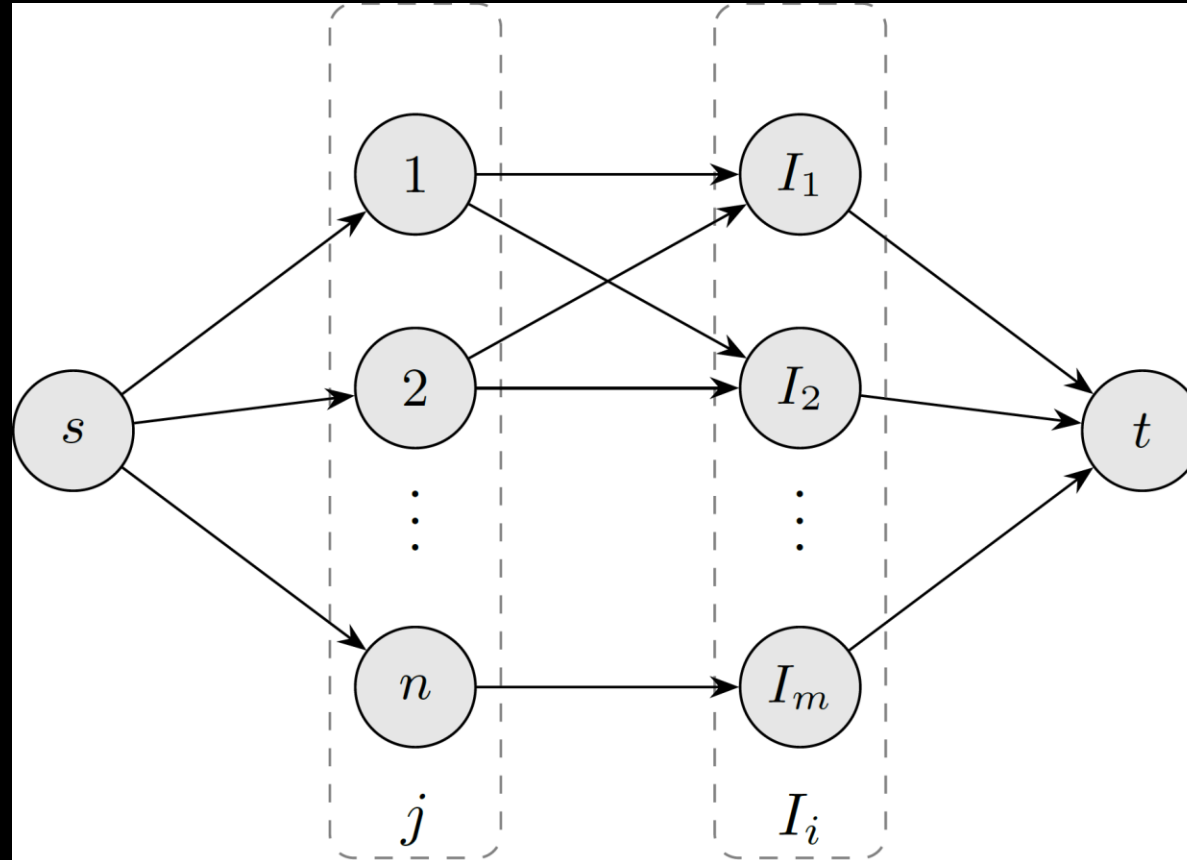
With speed-limits

- Jobs j :
 - Arrival times a_j
 - Departure times d_j
 - Work e_j
 - Maximum speed $p_j^{\max}$
- Decision variable:
 - Job schedules $s_j(t)$
- Objective function (energy):
 - $\int_t (\sum_j s_j(t))^2 dt$

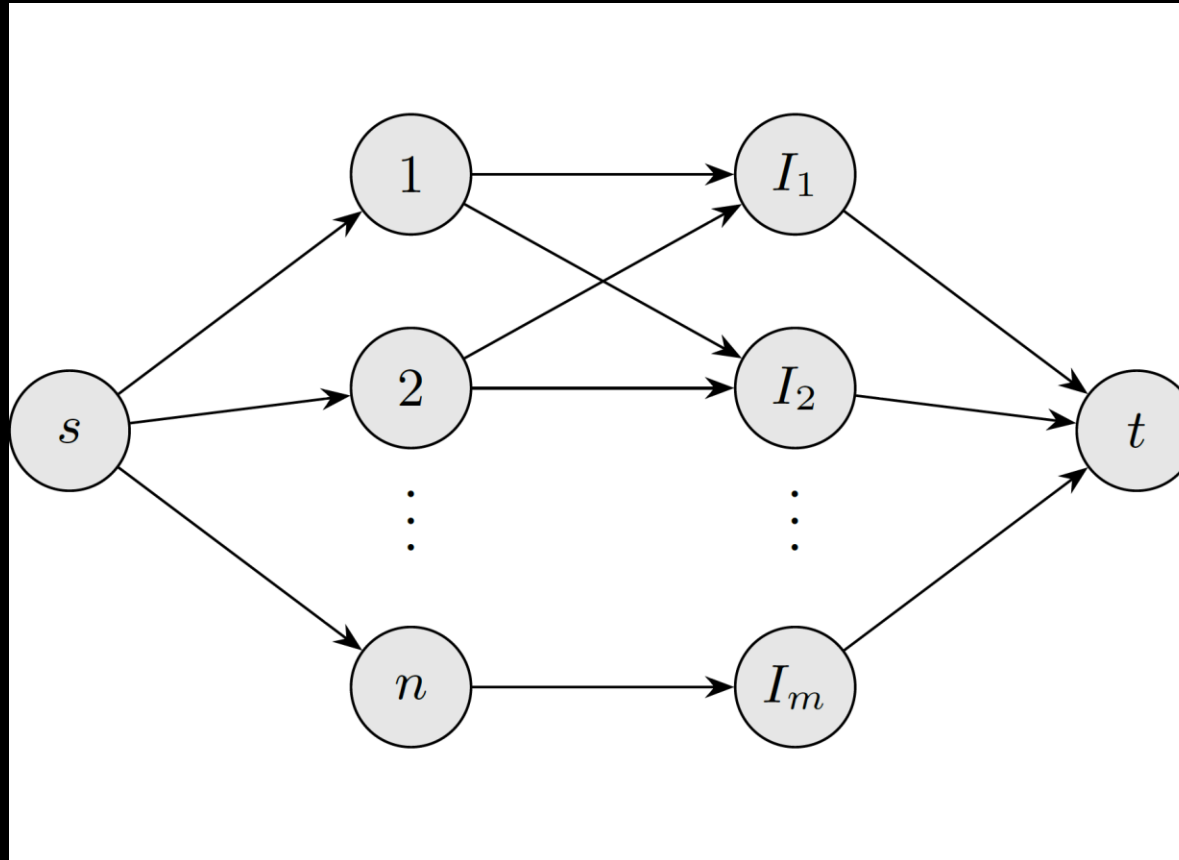
SPEED SCALING – FLOW MODEL



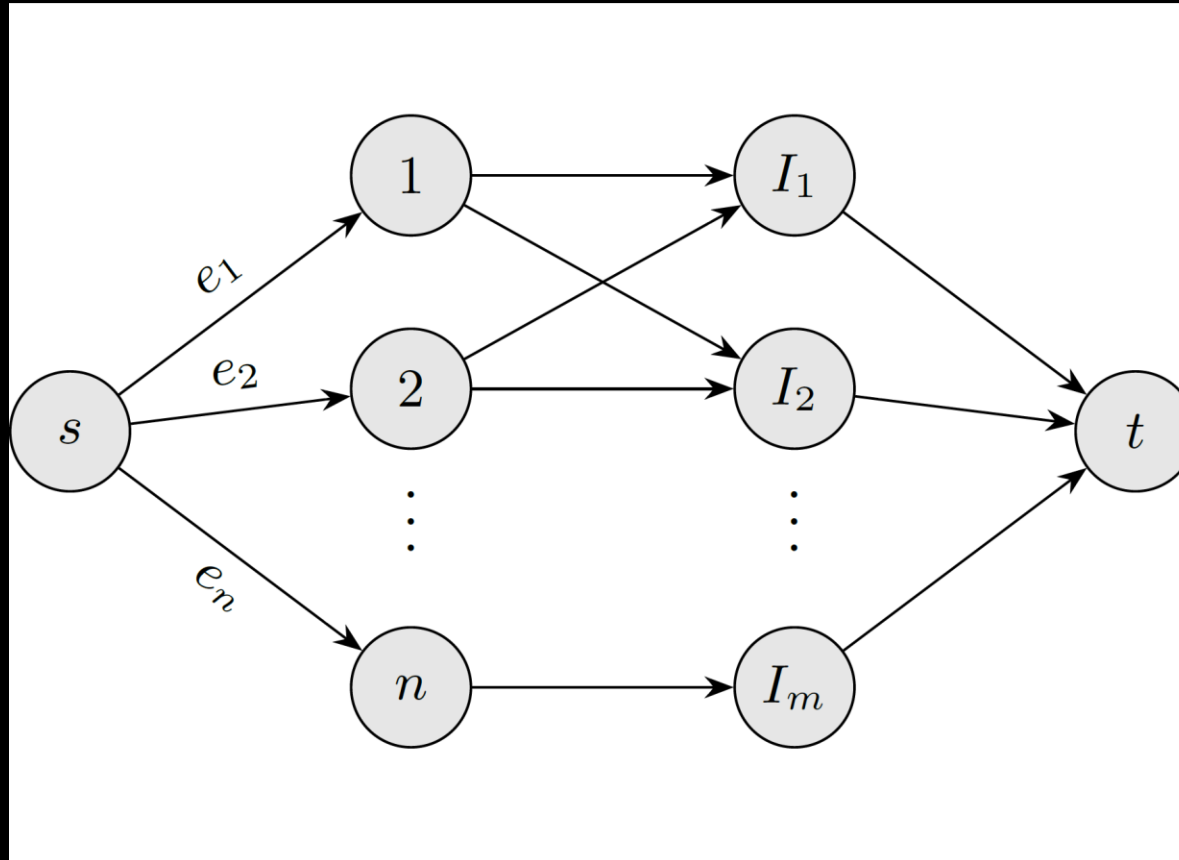
SPEED SCALING – FLOW MODEL



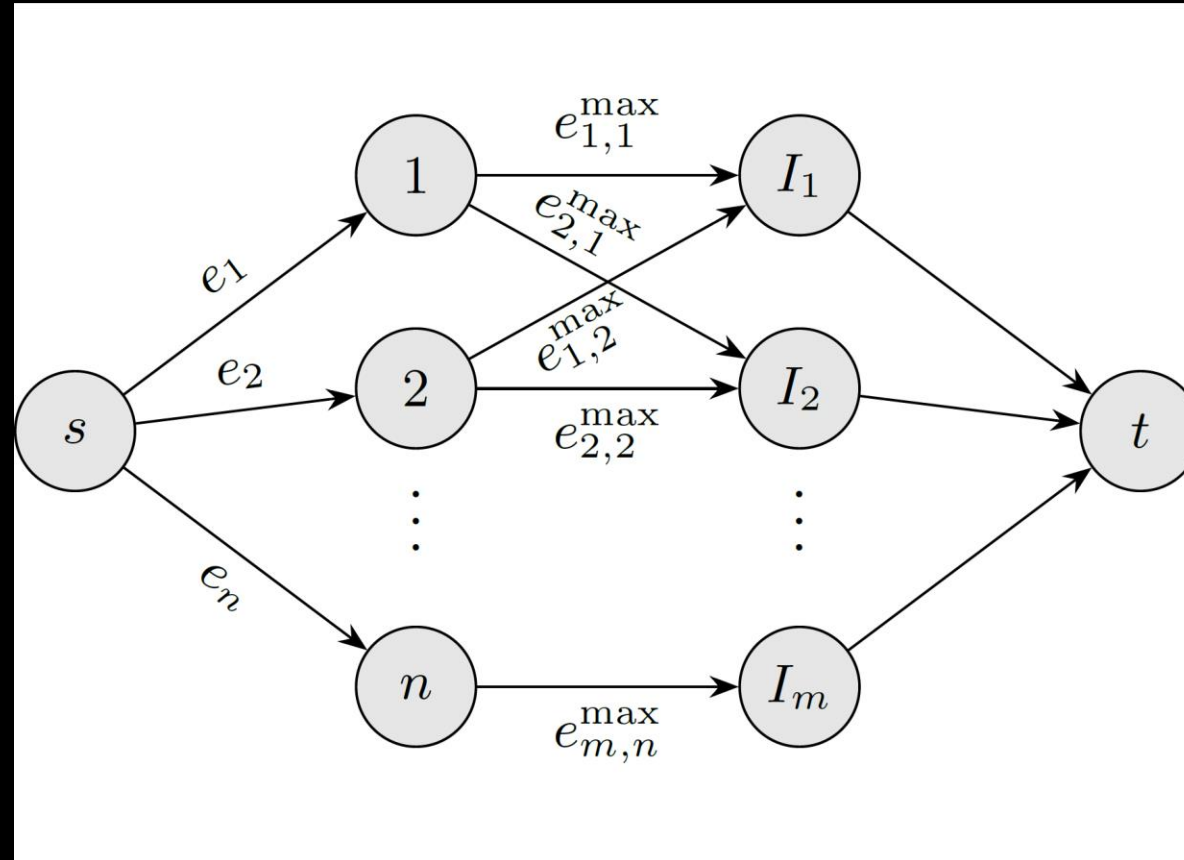
SPEED SCALING – FLOW MODEL



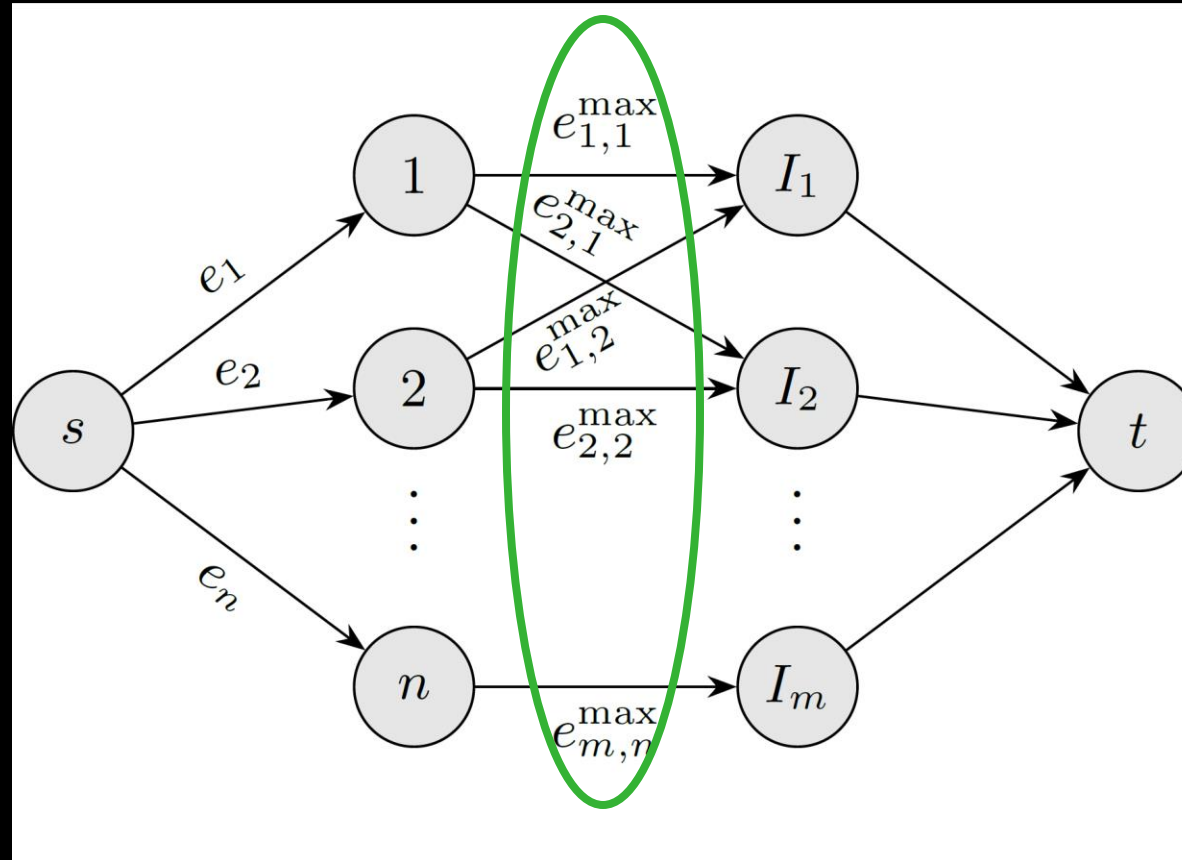
SPEED SCALING – FLOW MODEL



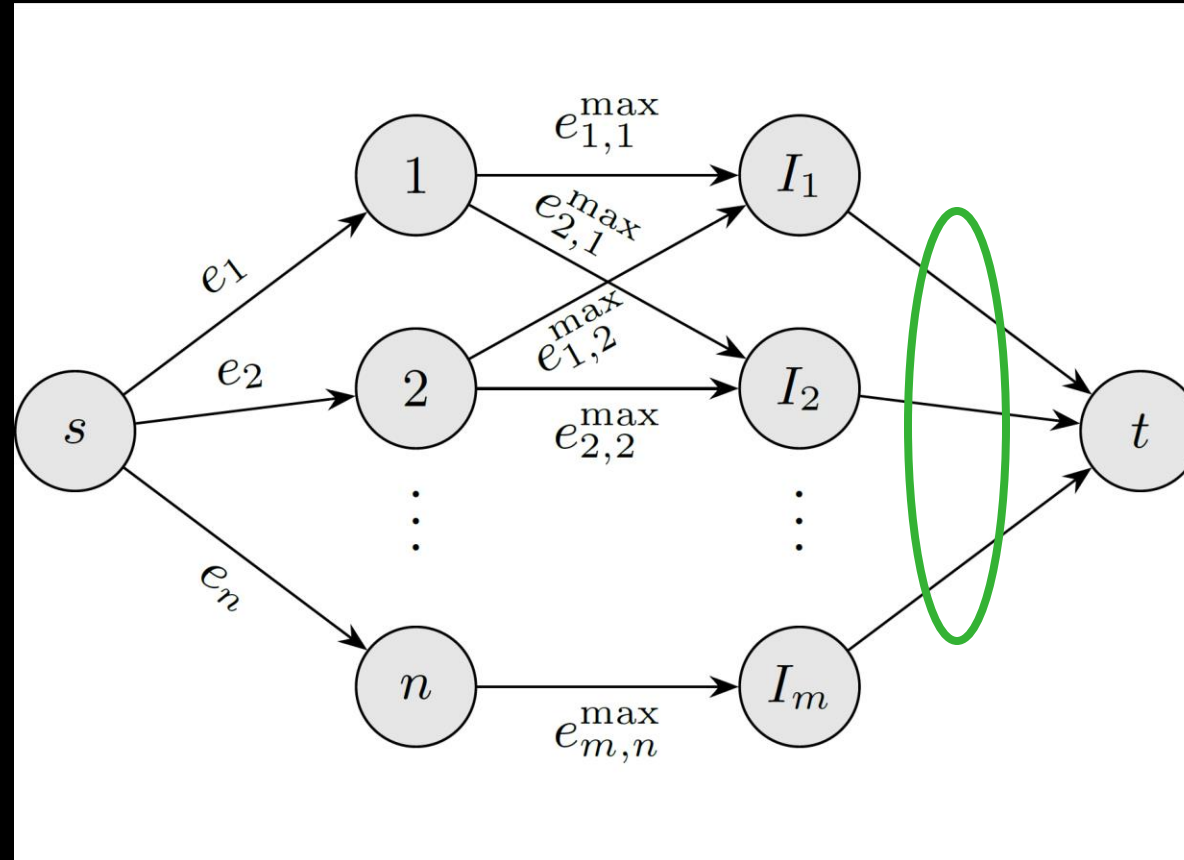
SPEED SCALING – FLOW MODEL



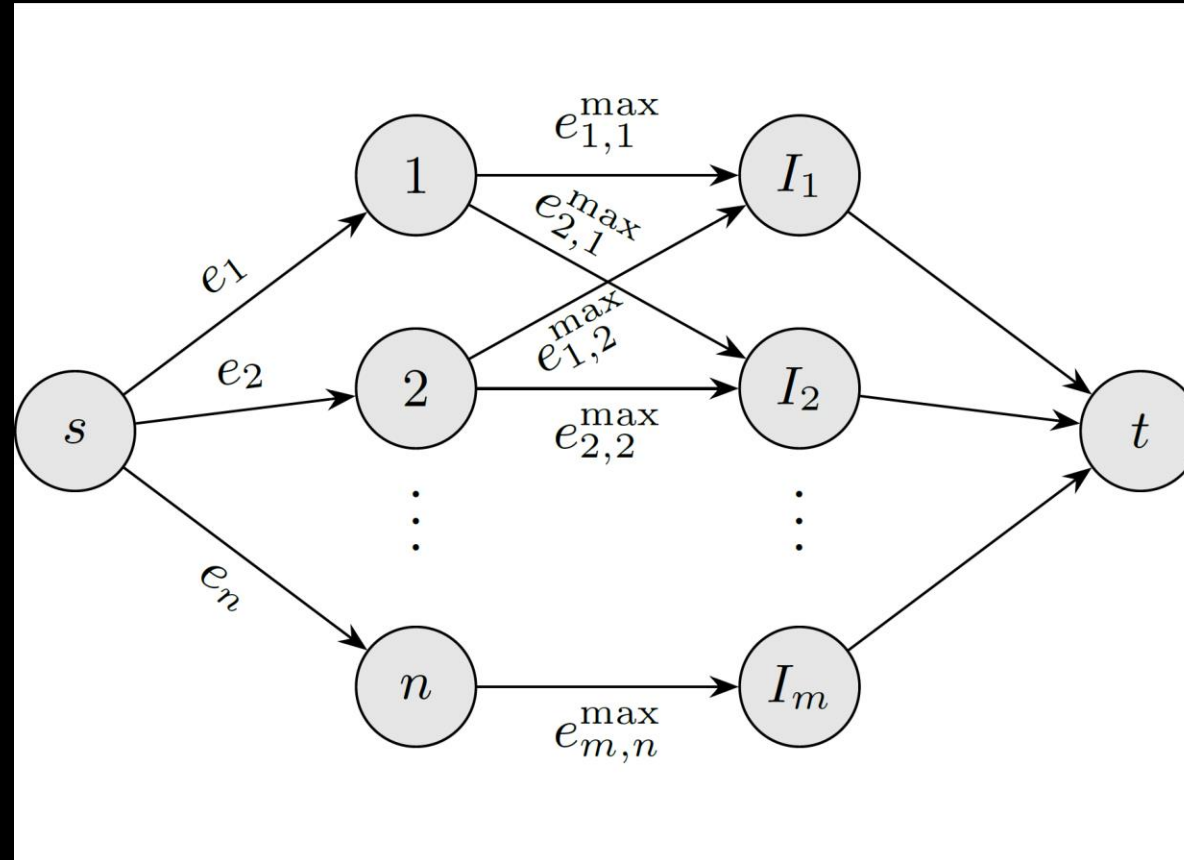
SPEED SCALING – FLOW MODEL

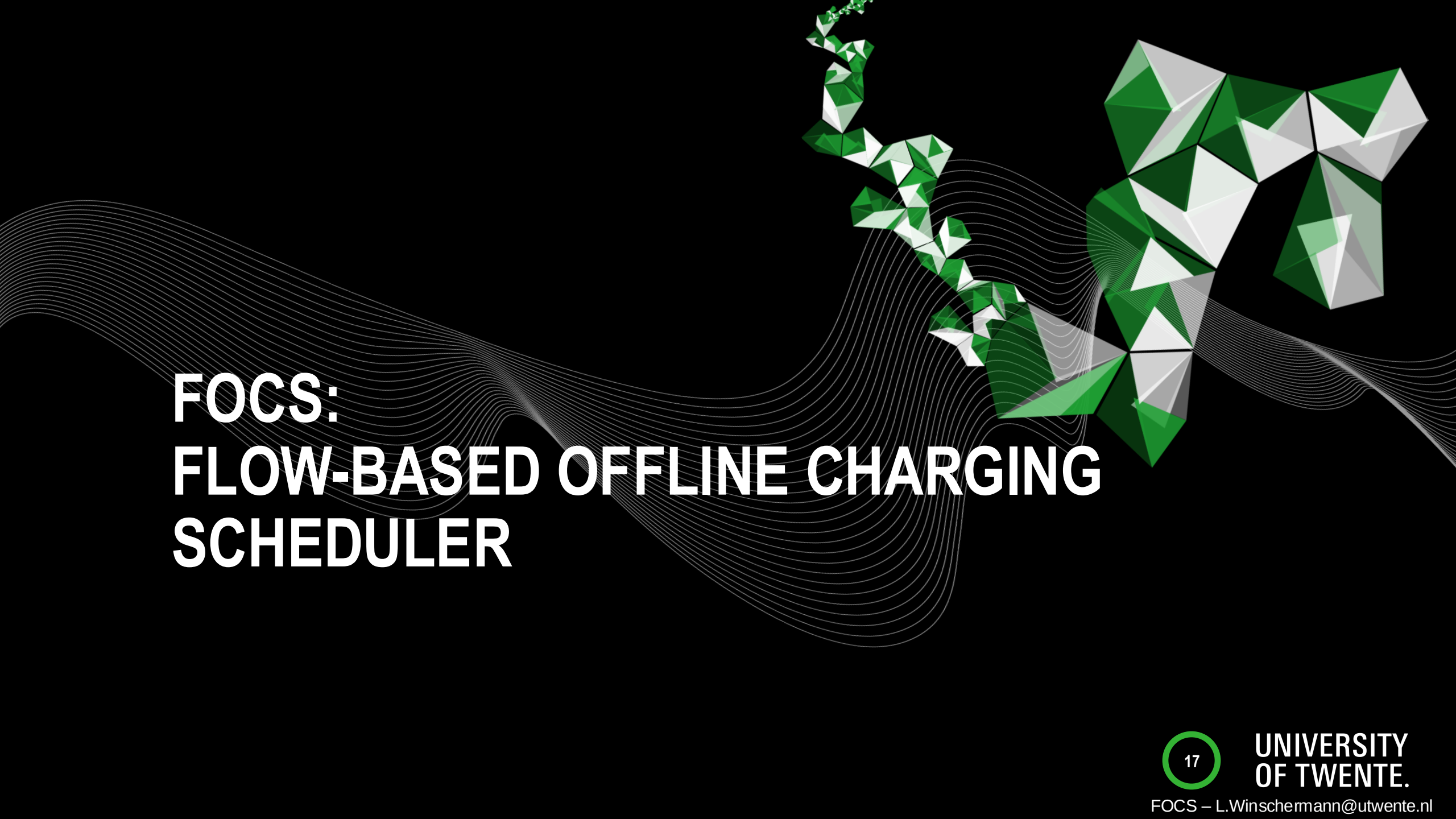


SPEED SCALING – FLOW MODEL



SPEED SCALING – FLOW MODEL





FOCS: FLOW-BASED OFFLINE CHARGING SCHEDULER



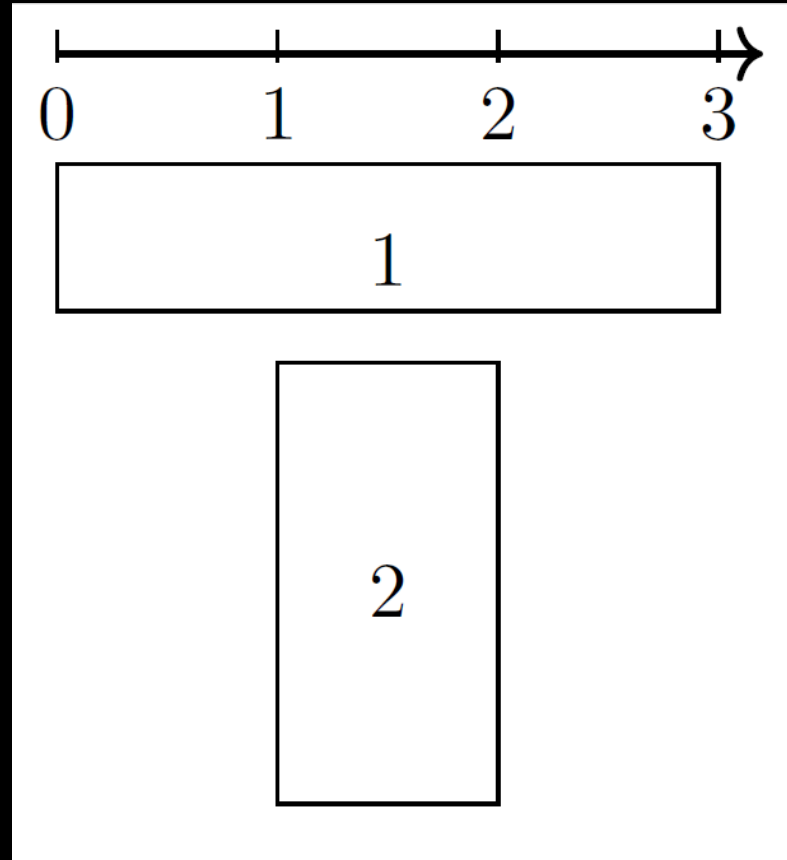
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FOCS – ILLUSTRATIVE EXAMPLE

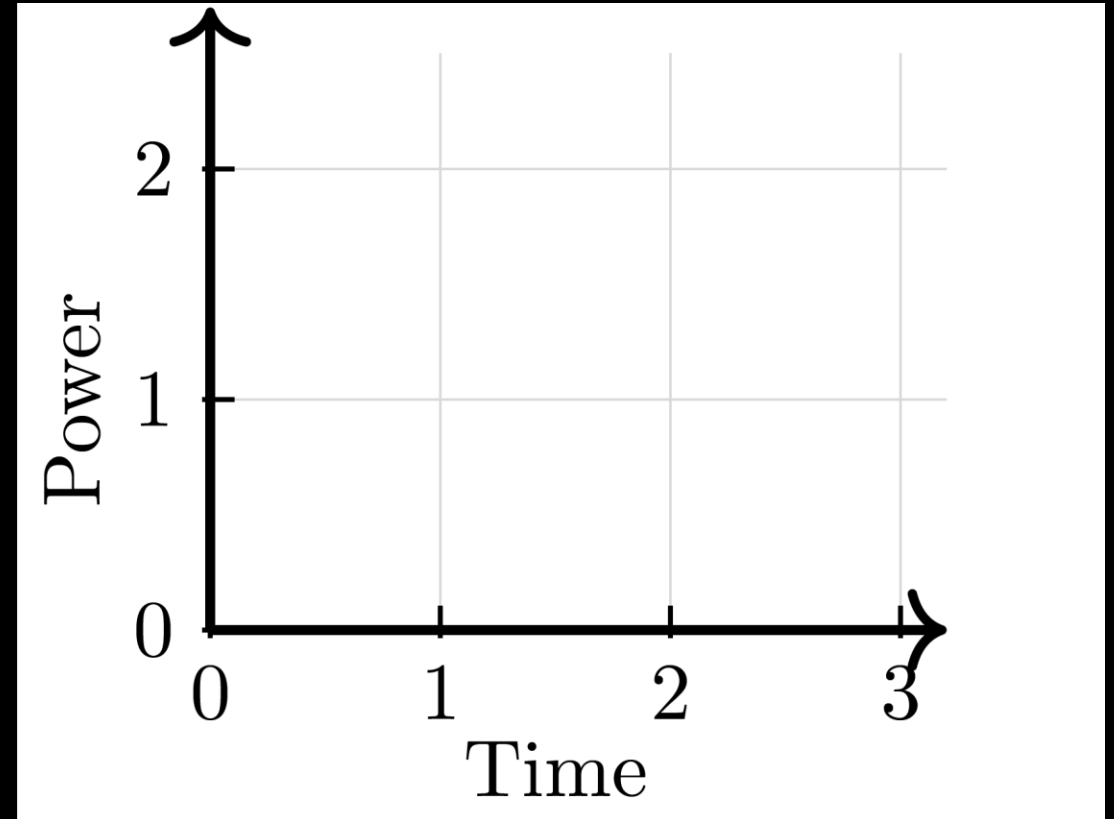
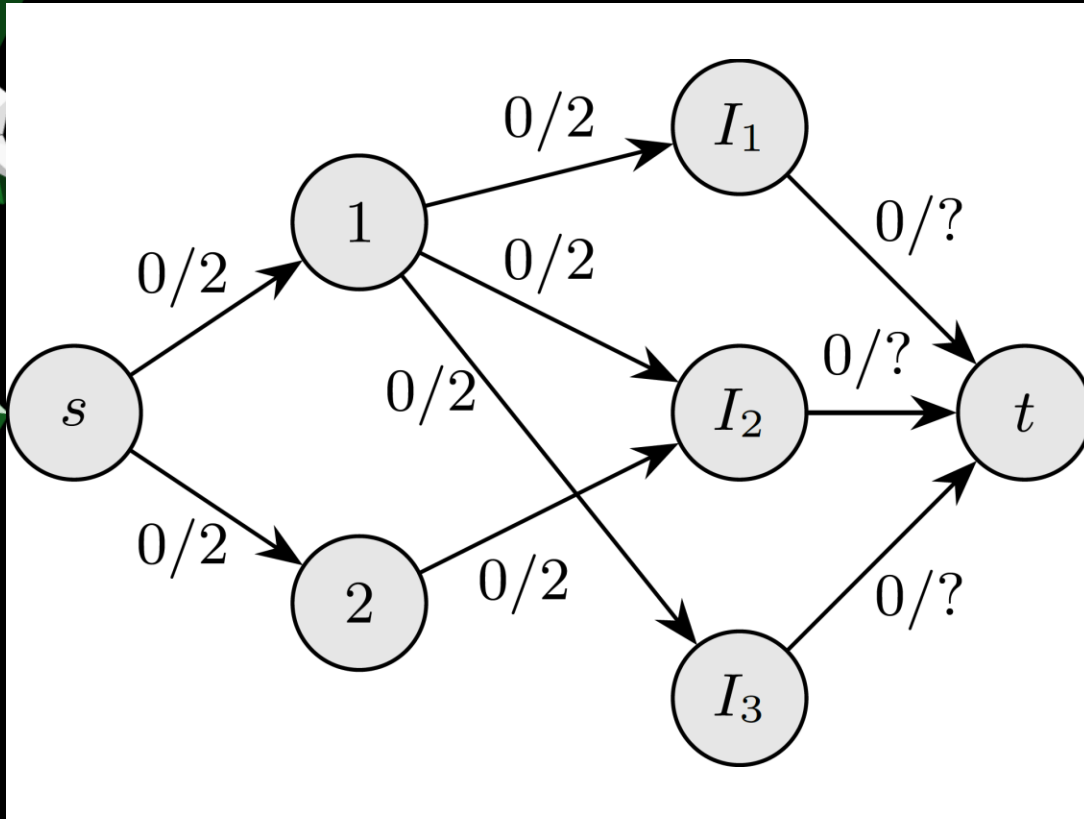


j	a_j	d_j	e_j	$p_j^{\max}$
1	0	3	2	2
2	1	2	2	2



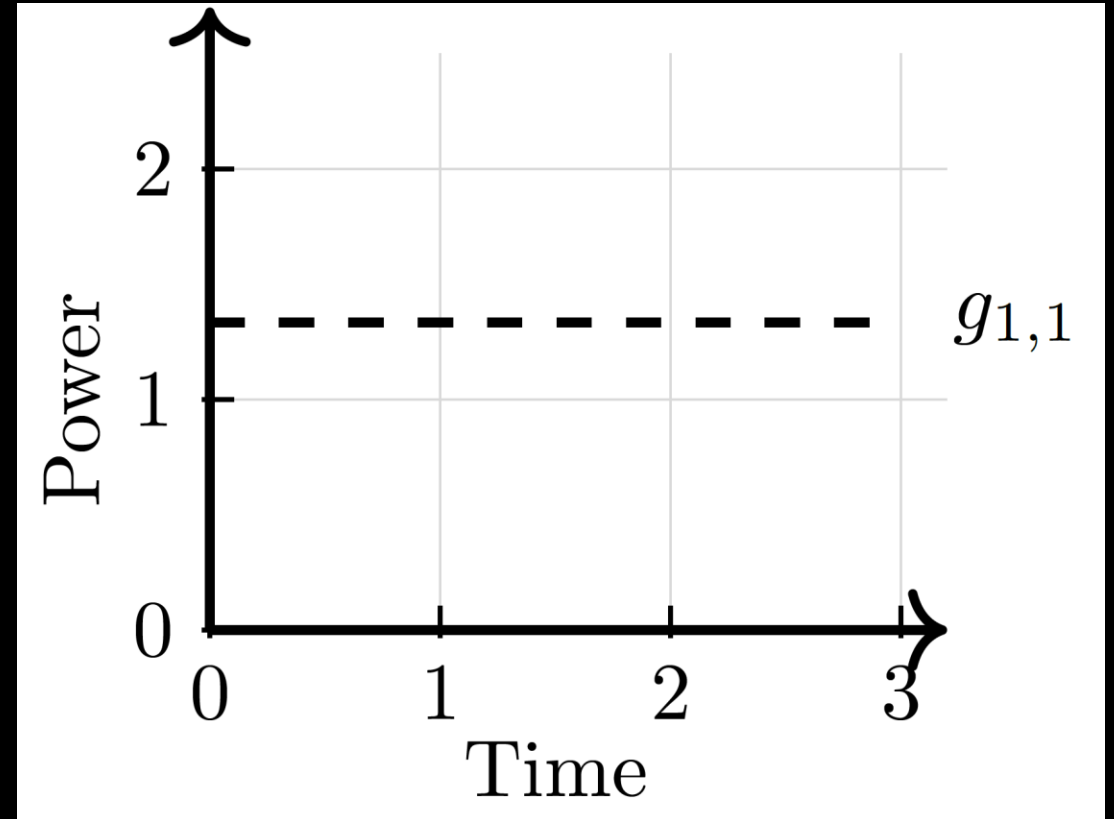
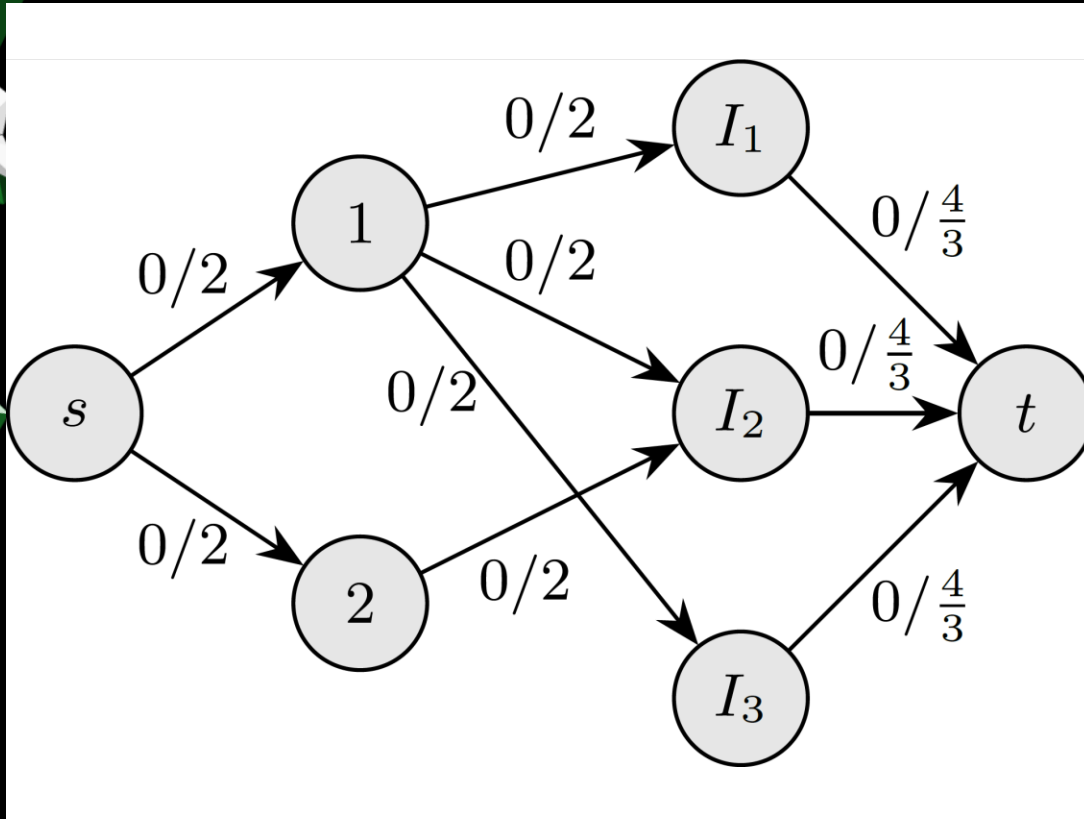
FOCS – ILLUSTRATIVE EXAMPLE

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1	0	3	2	2
2	1	2	2	2



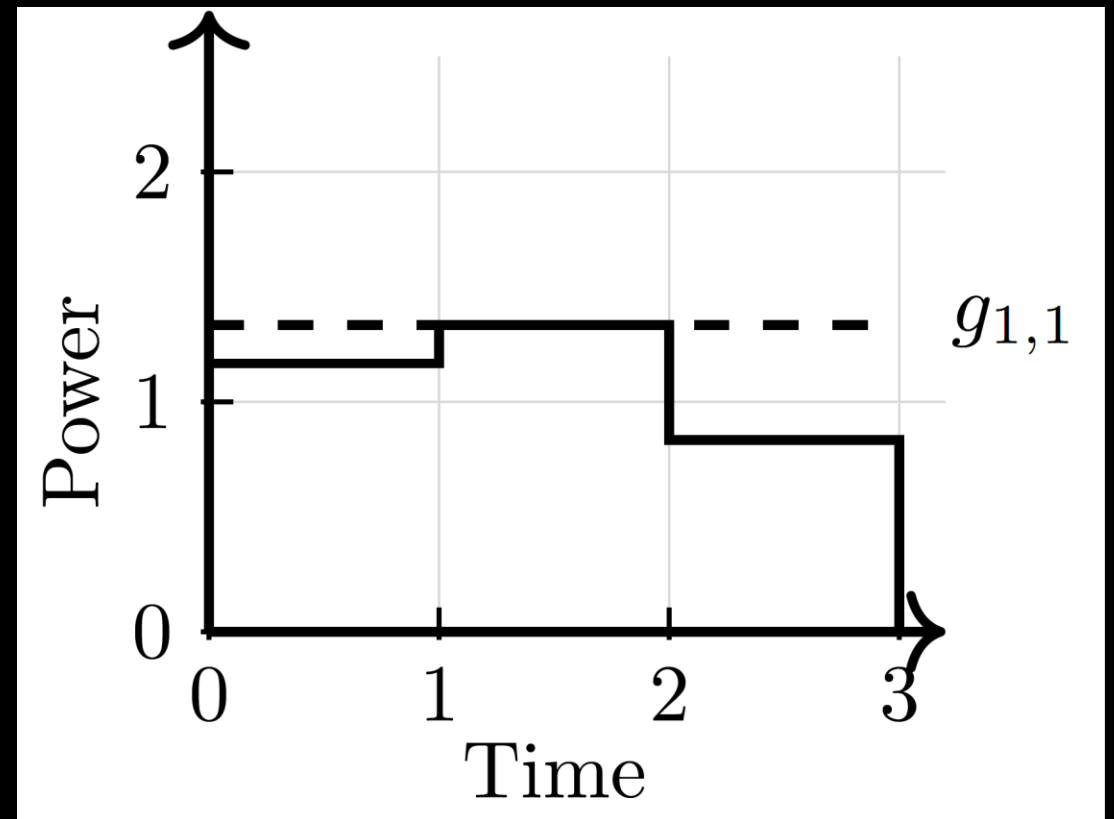
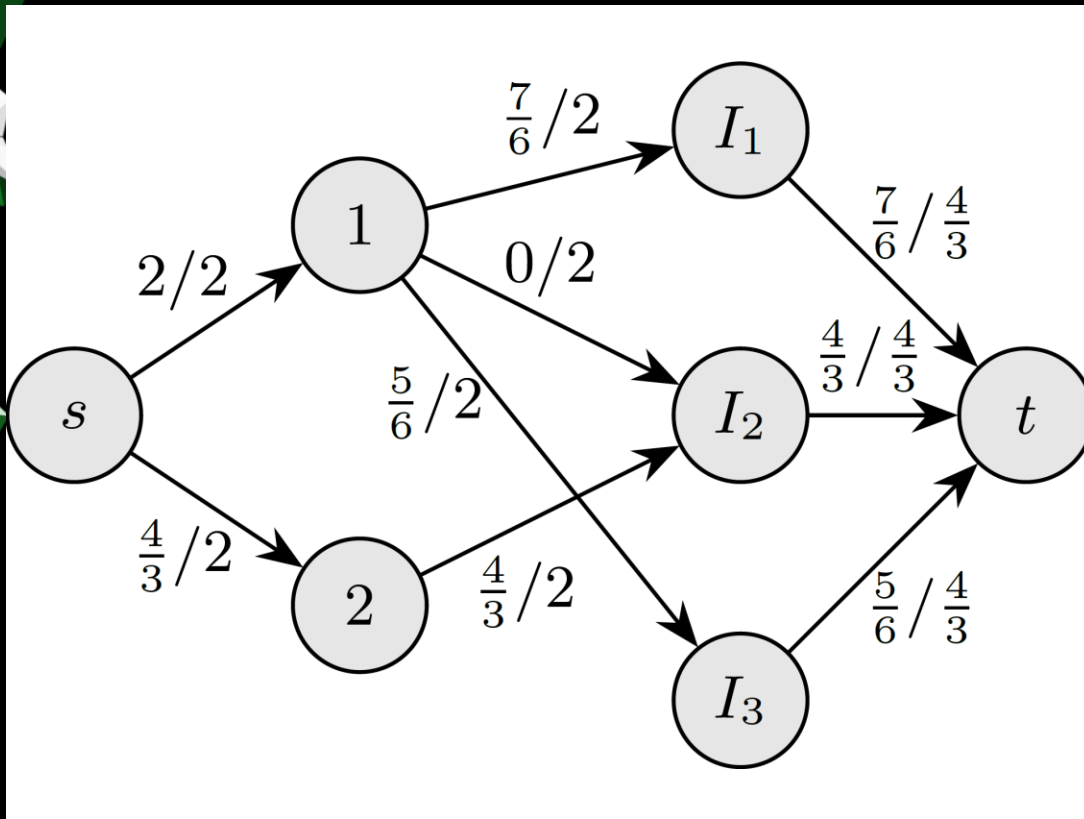
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1	0	3	2	2
2	1	2	2	2



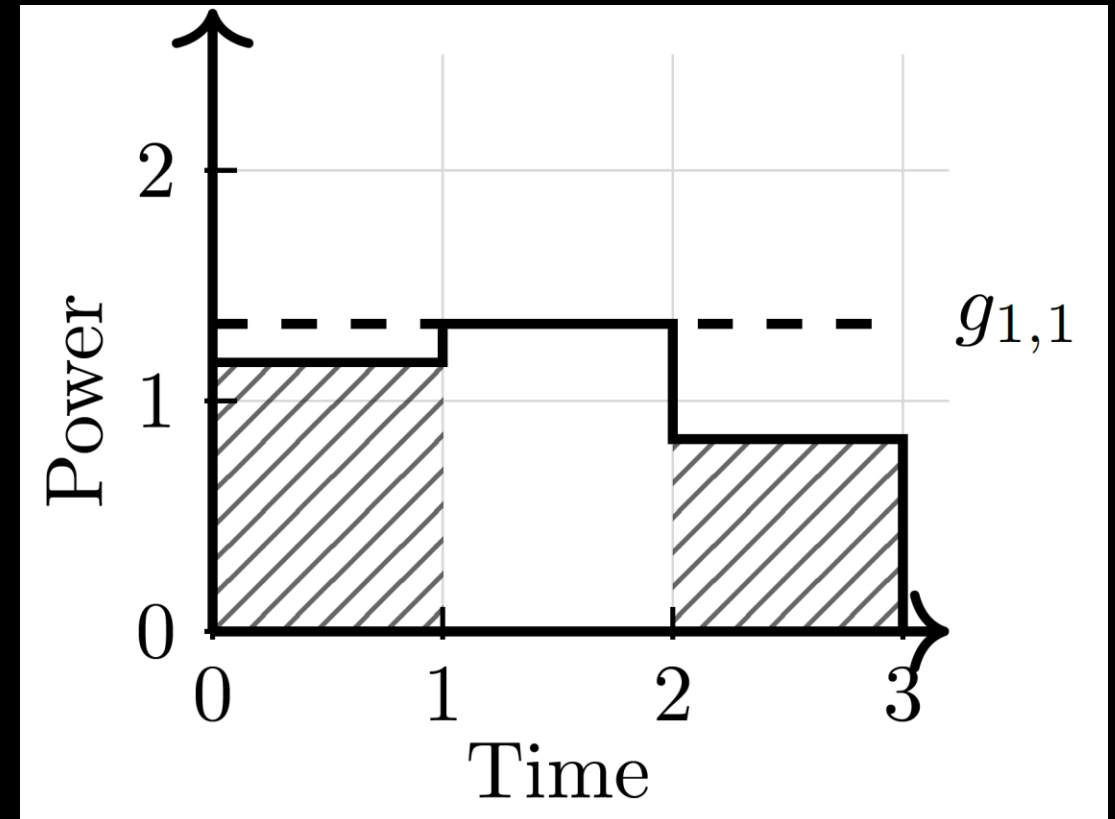
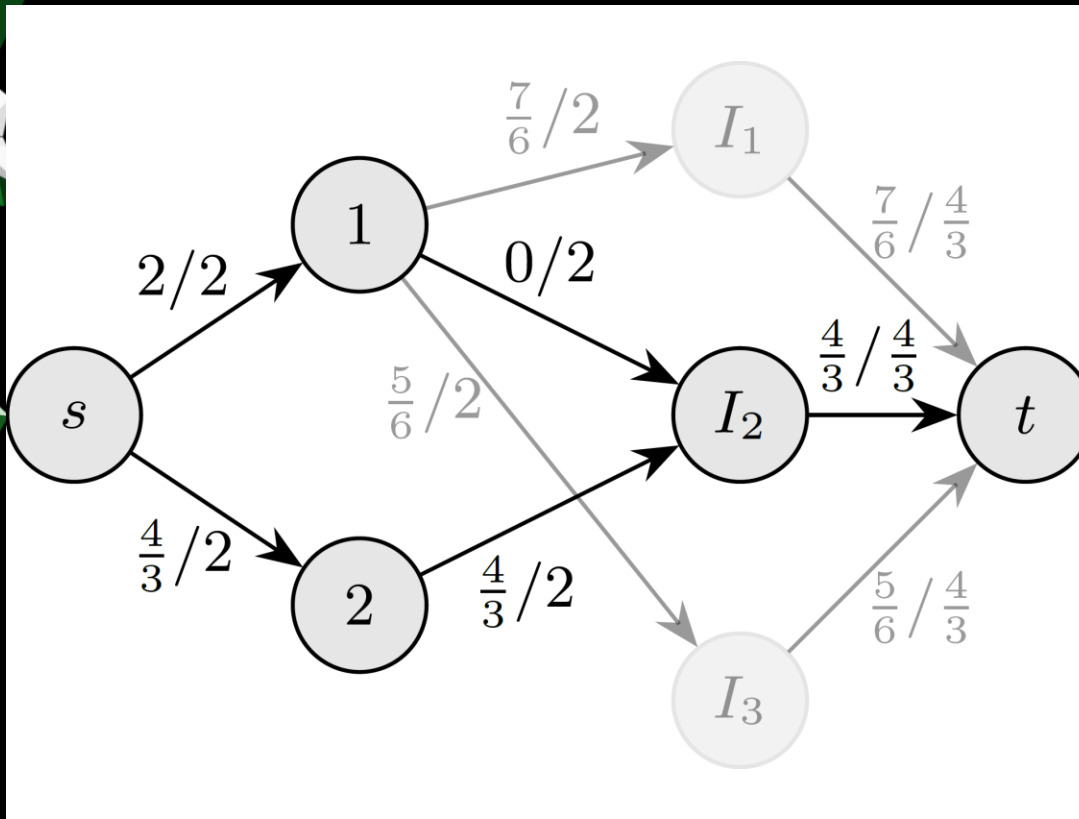
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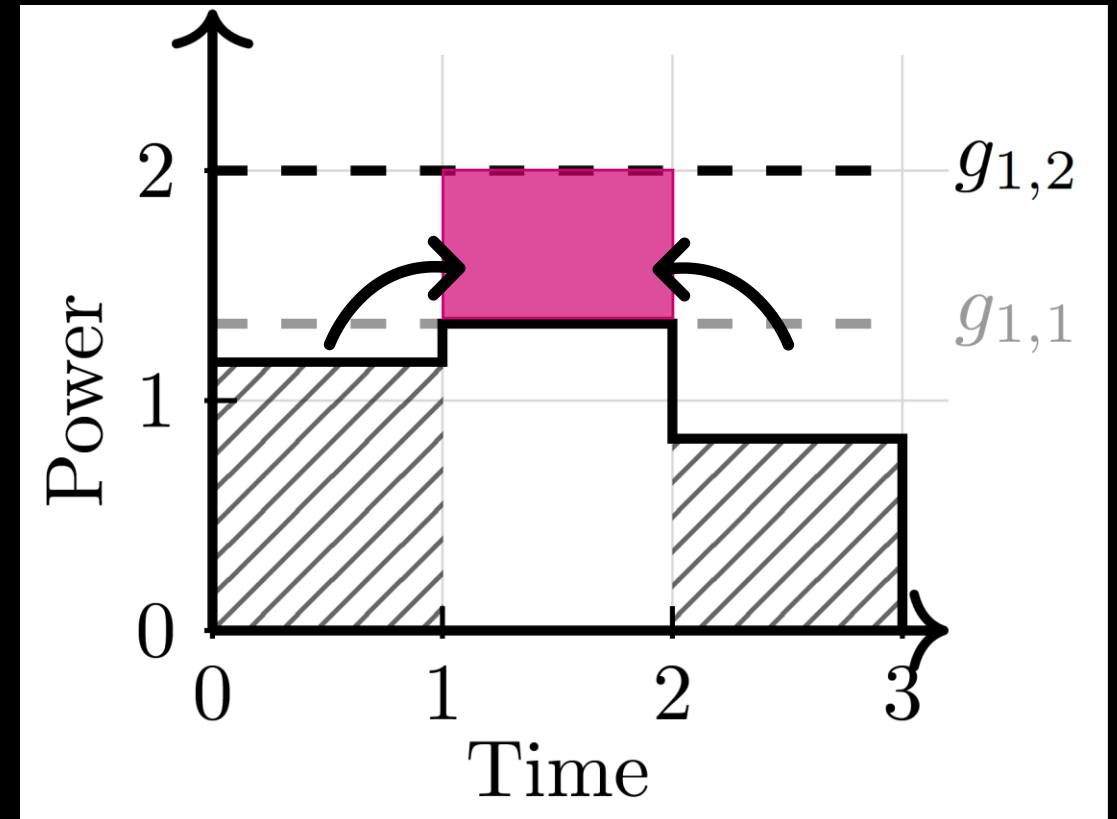
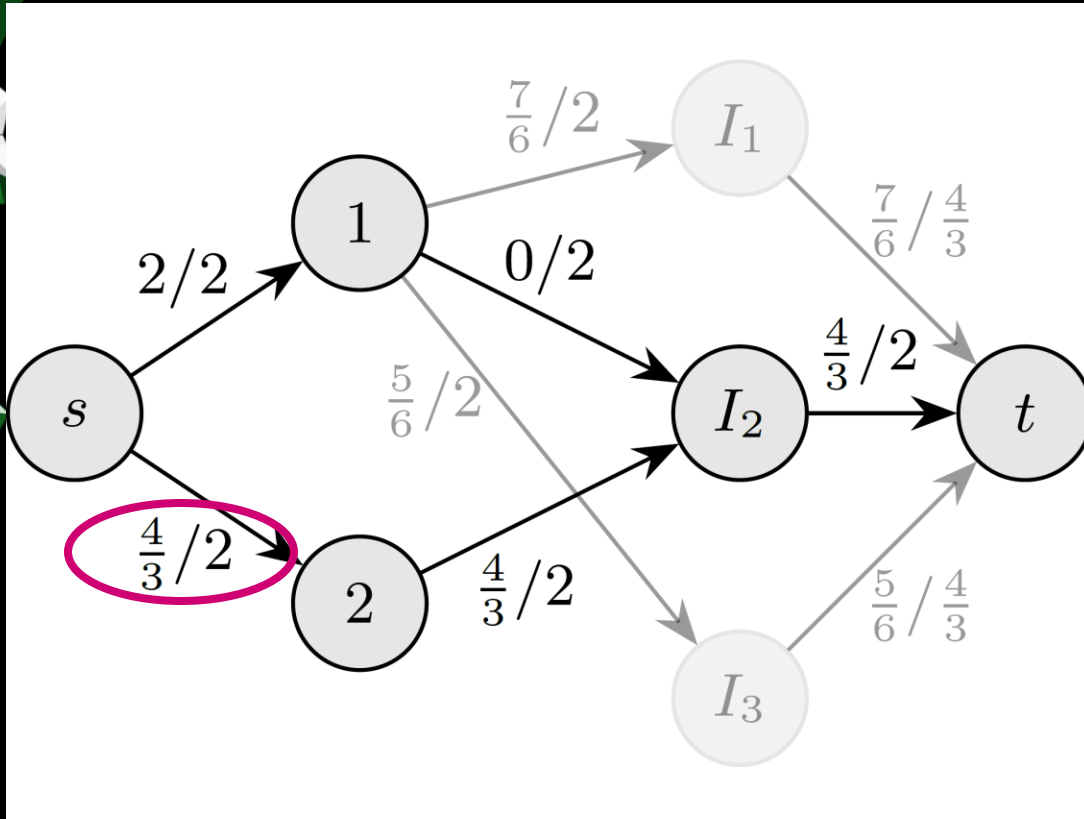
FOCS – ILLUSTRATIVE EXAMPLE

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1	0	3	2	2
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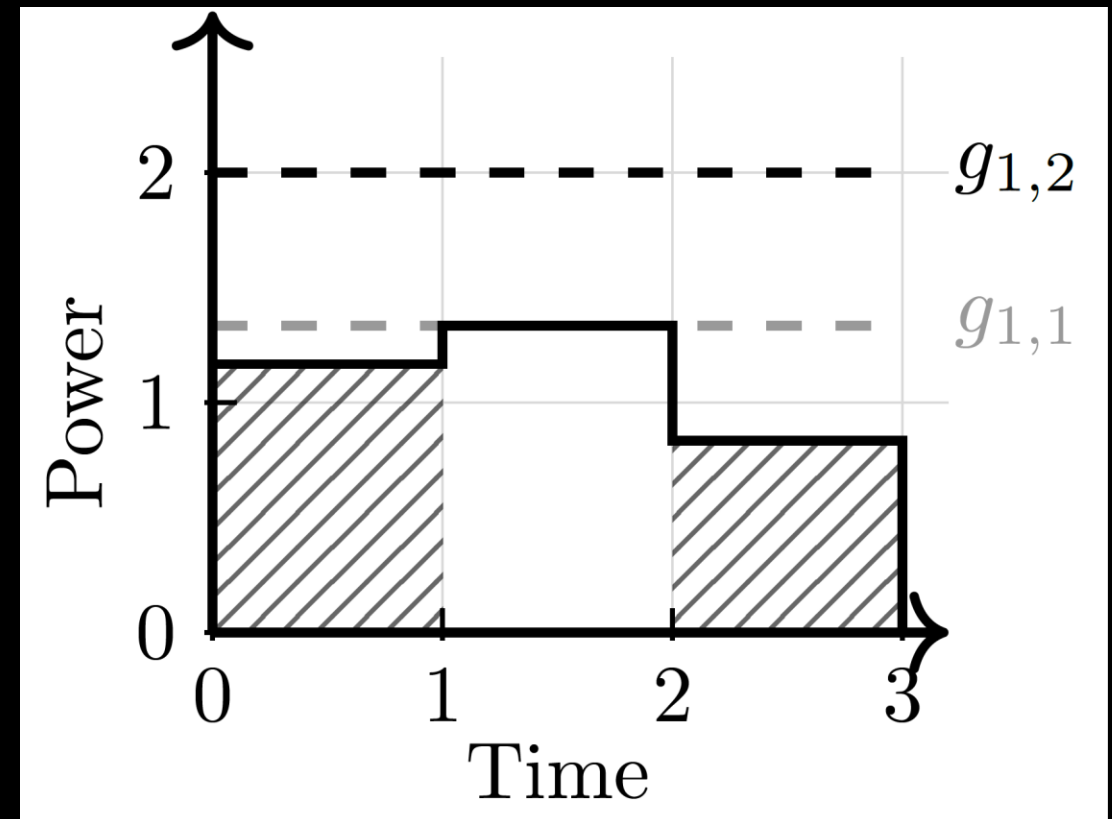
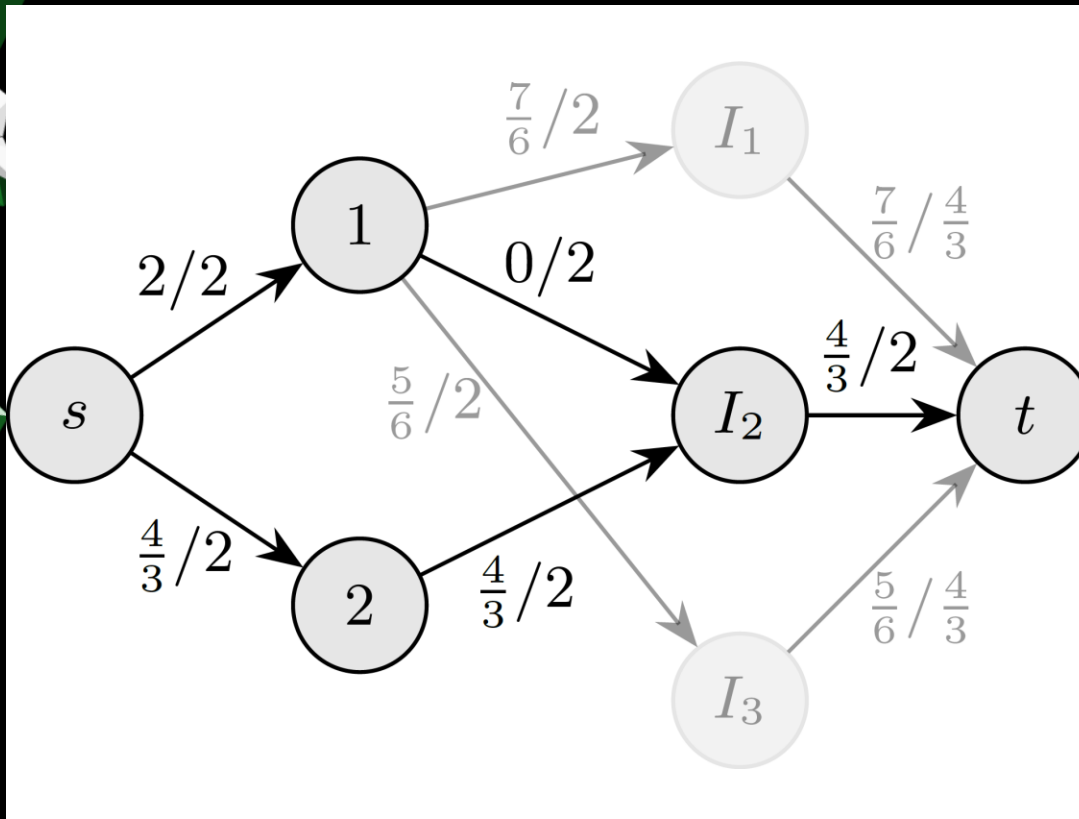
FOCS – ILLUSTRATIVE EXAMPLE

j	a_j	d_j	e_j	$p_j^{\max}$
1	0	3	2	2
2	1	2	2	2



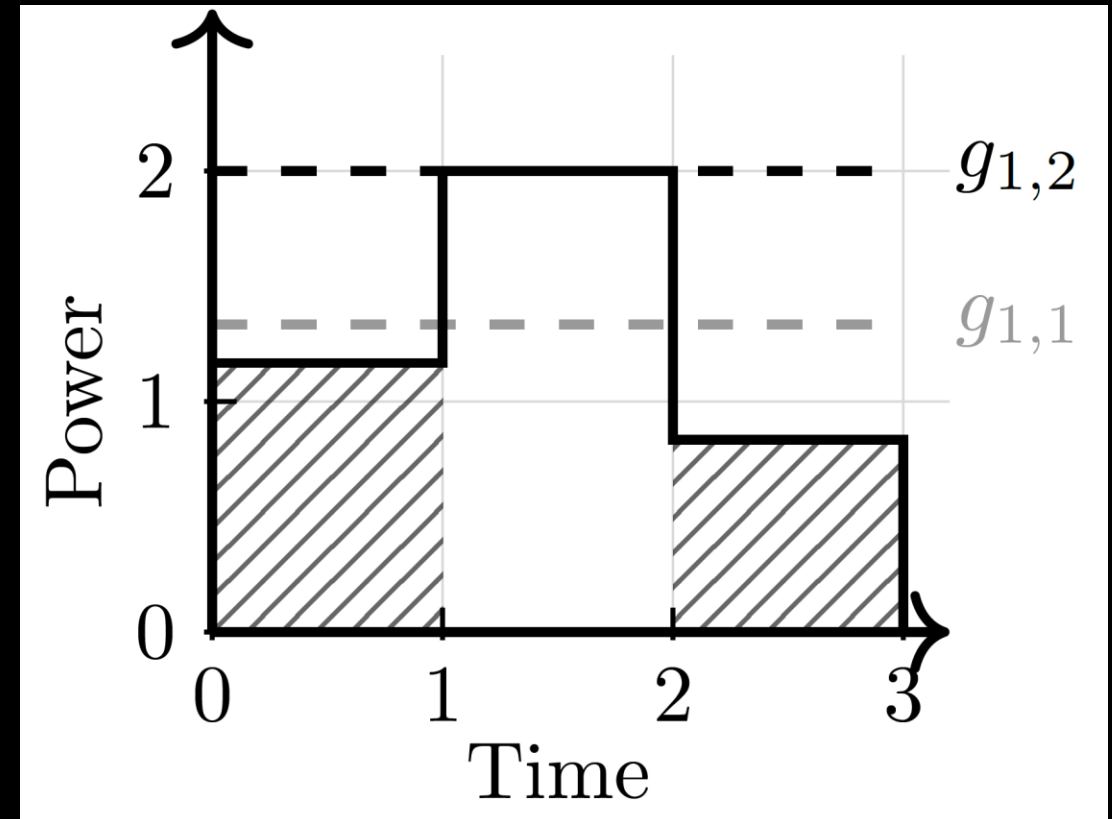
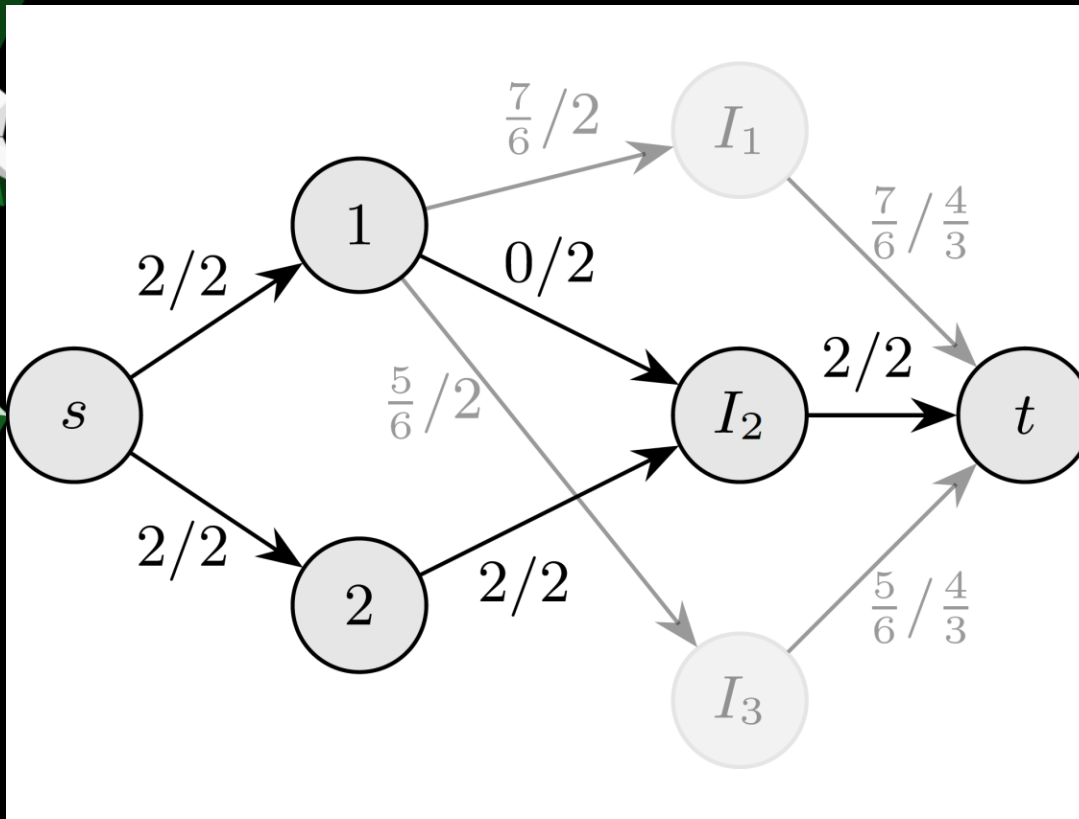
FOCS – ILLUSTRATIVE EXAMPLE

j	a_j	d_j	e_j	$p_j^{\max}$
1	0	3	2	2
2	1	2	2	2



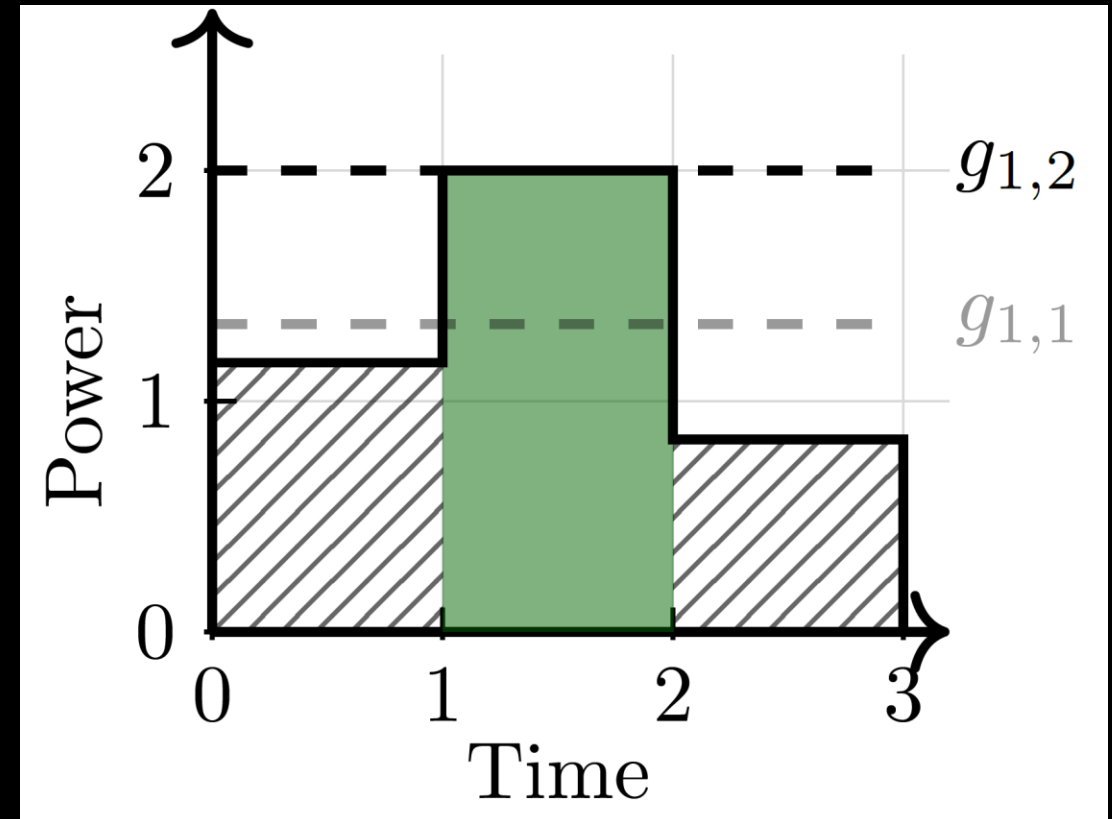
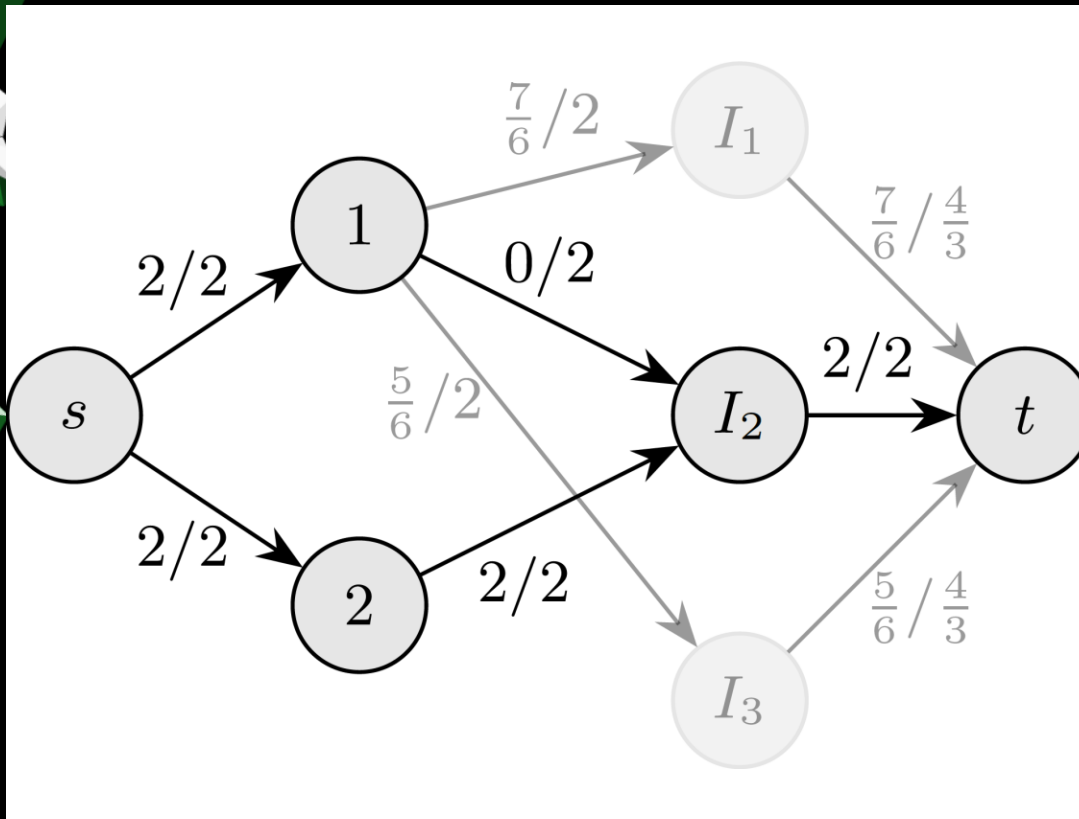
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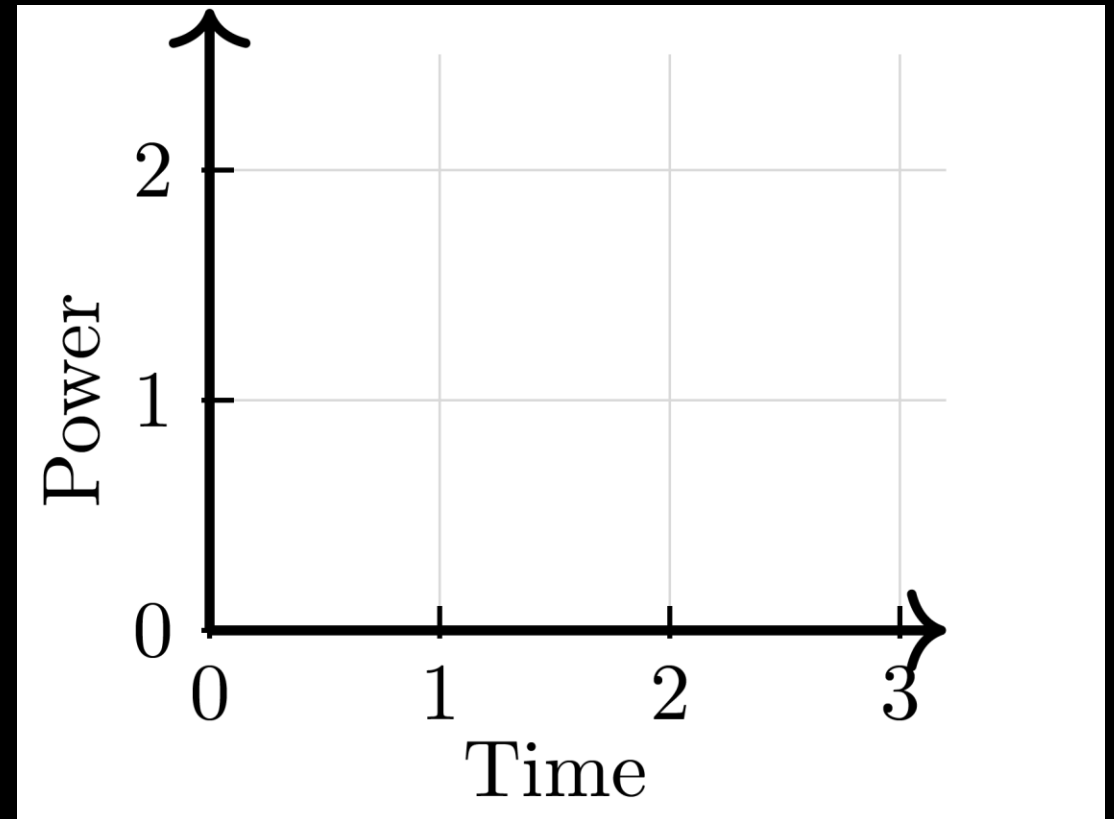
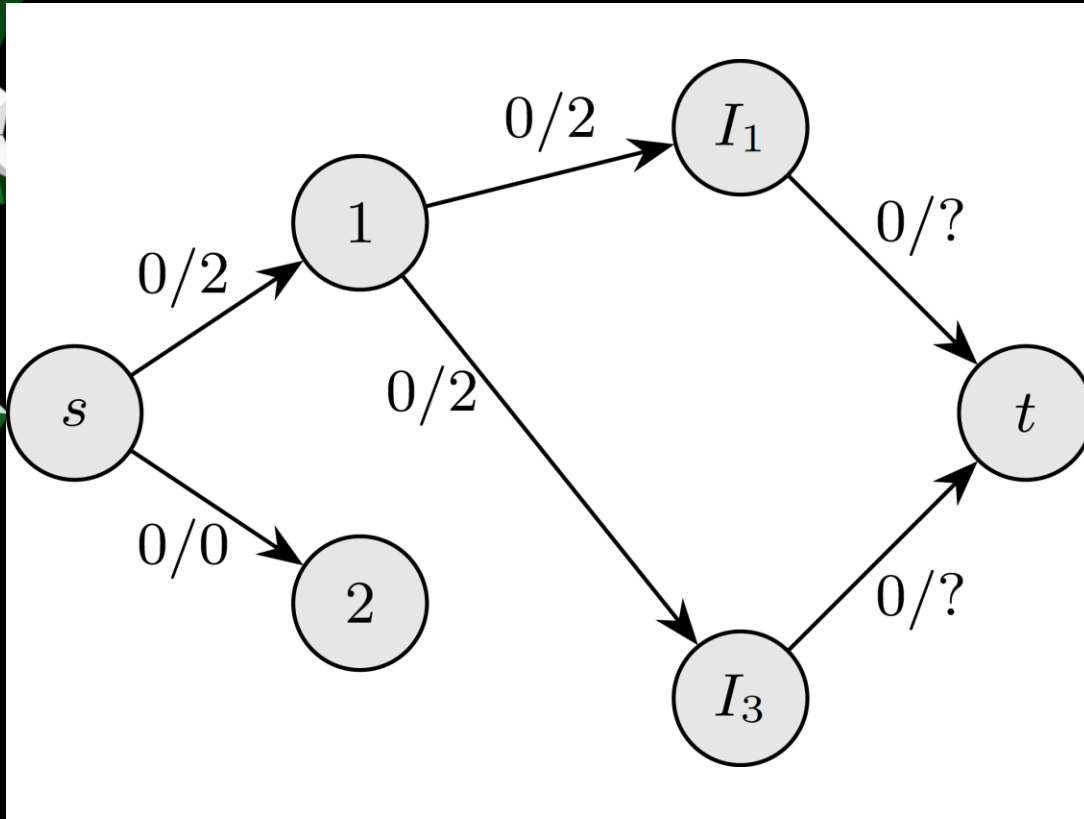
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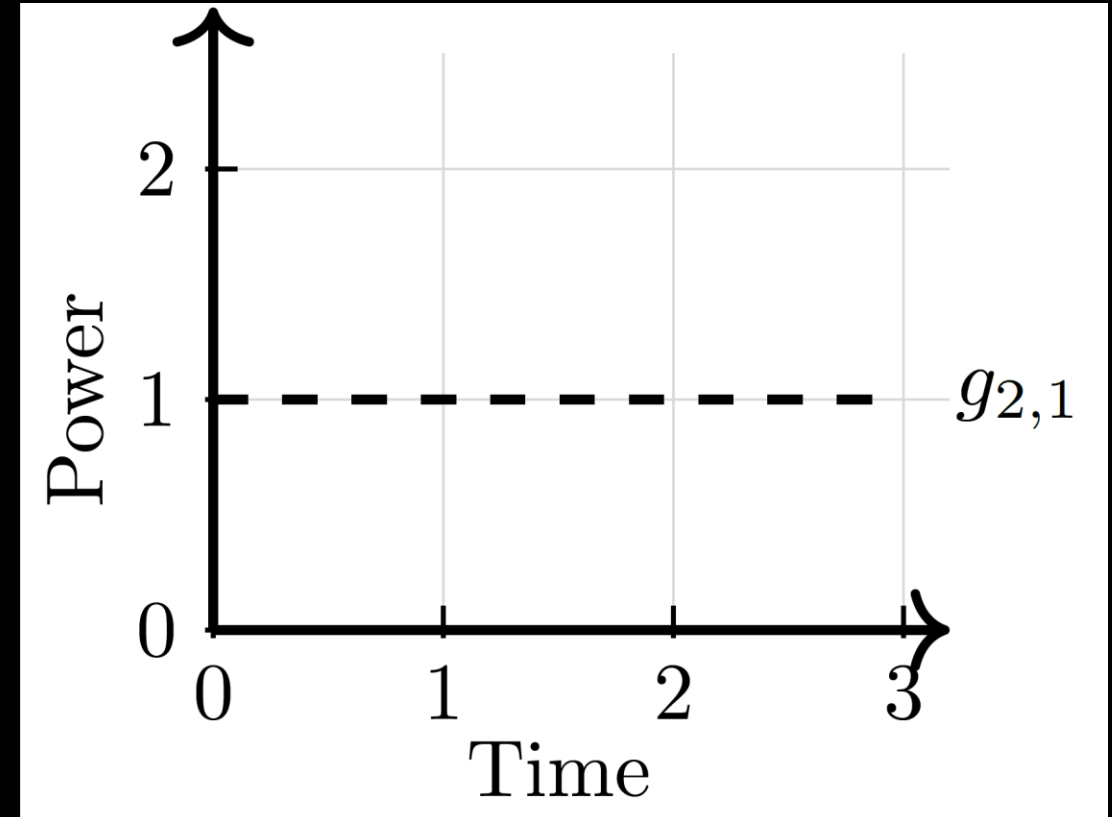
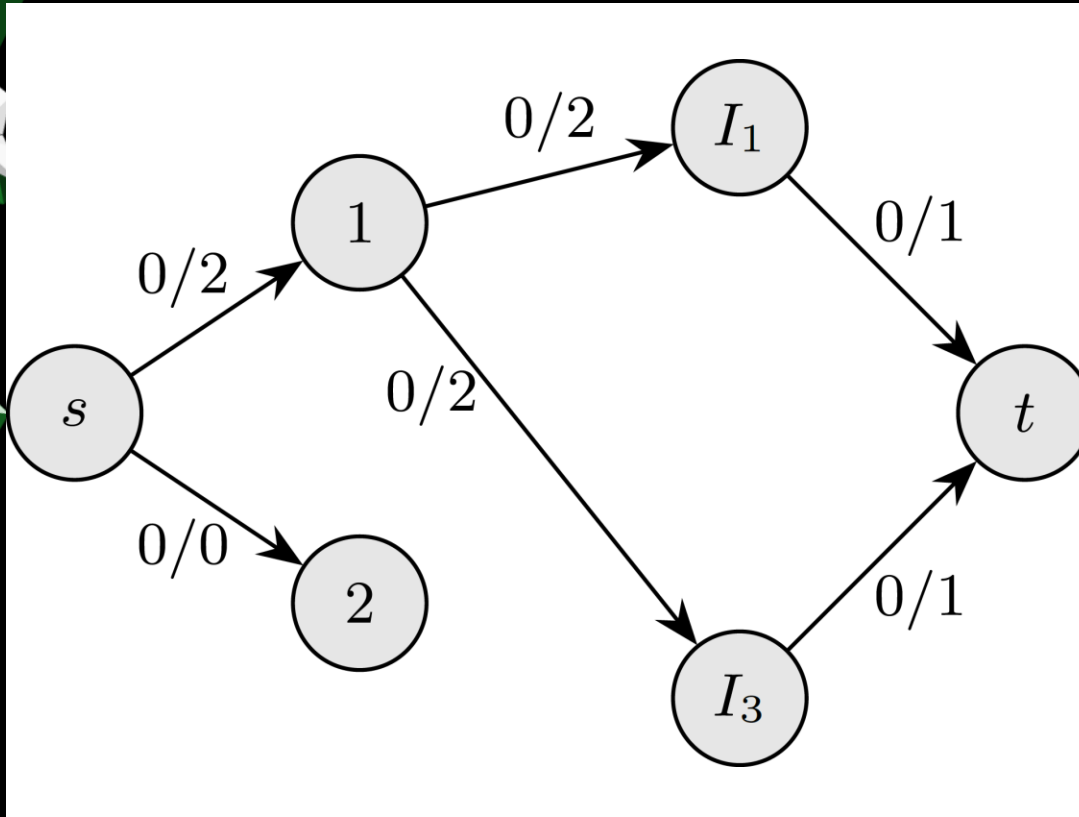
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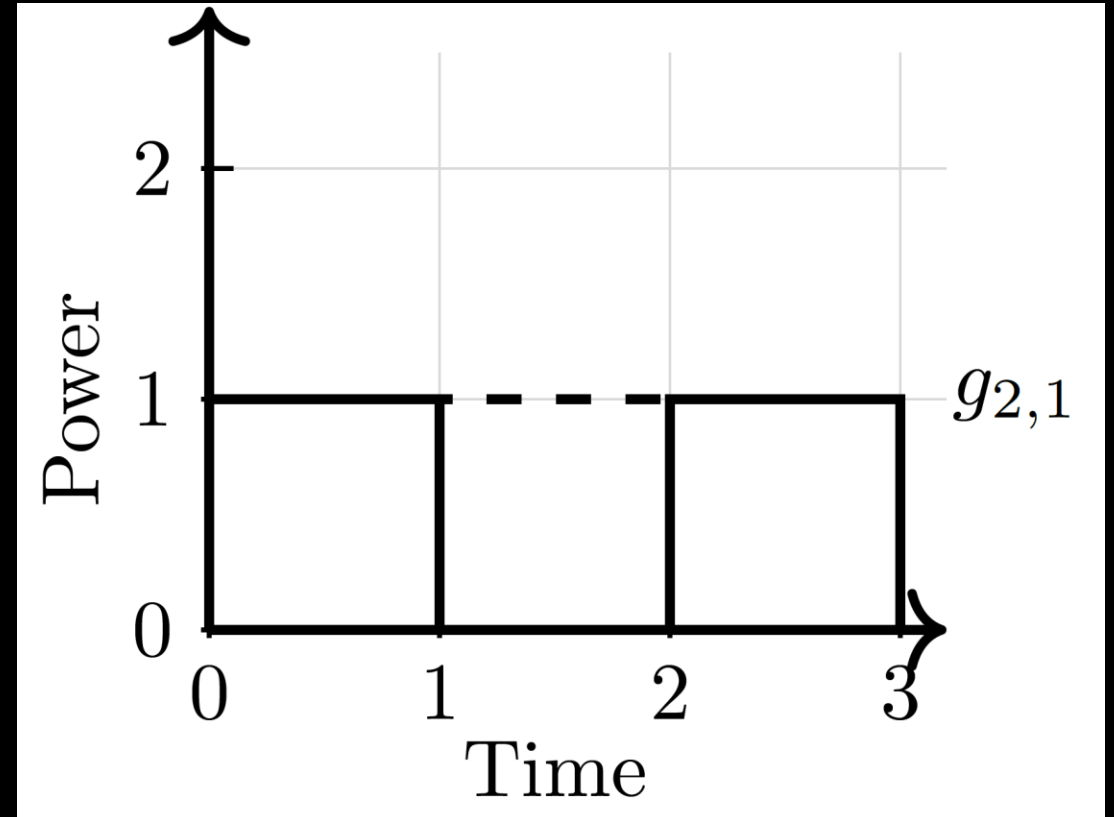
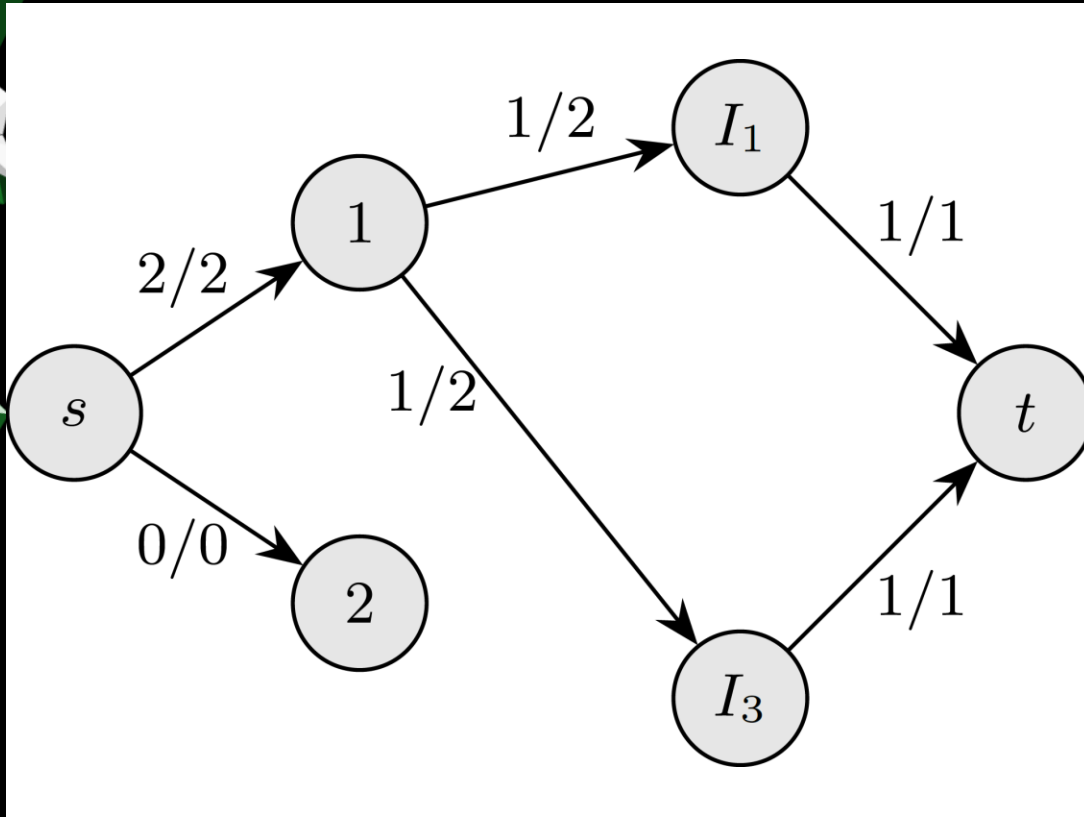
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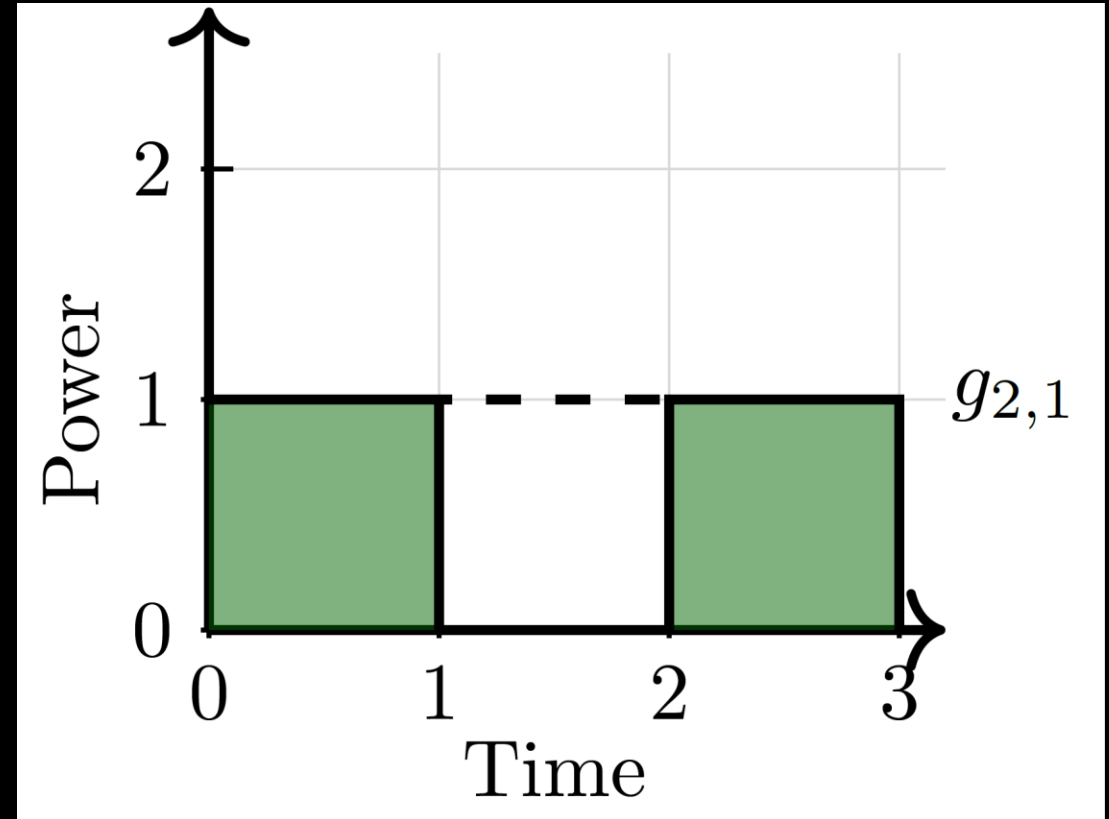
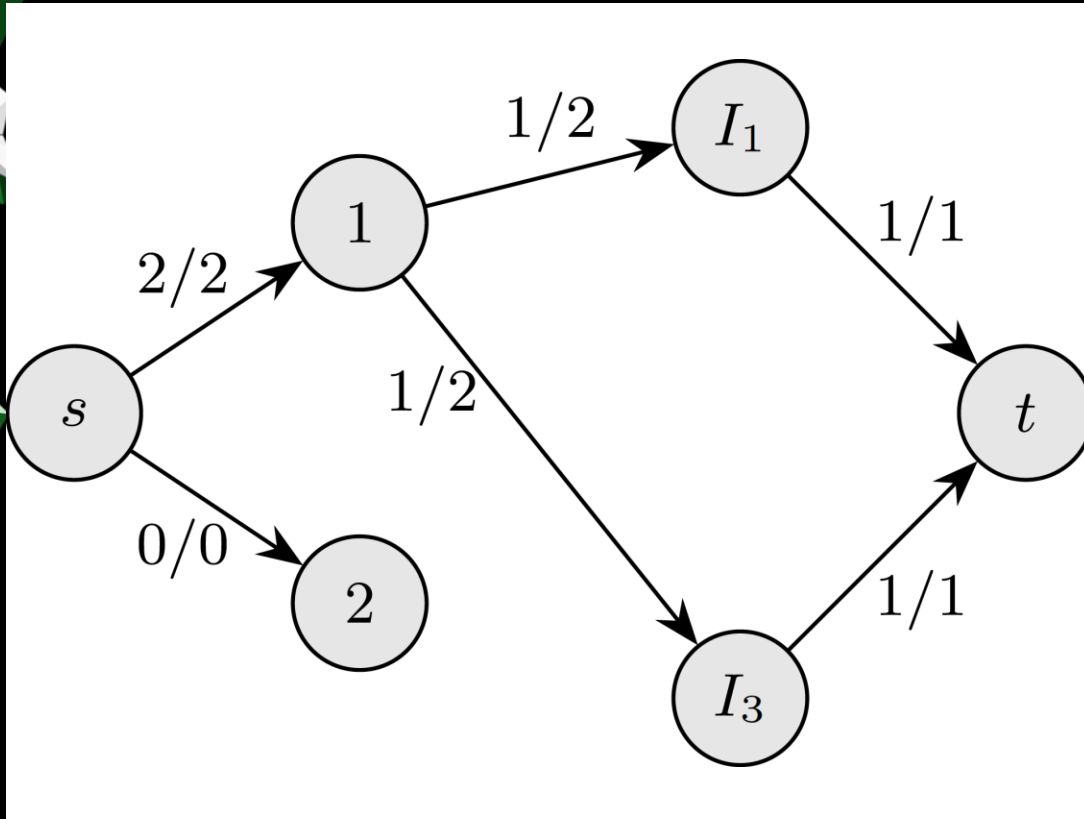
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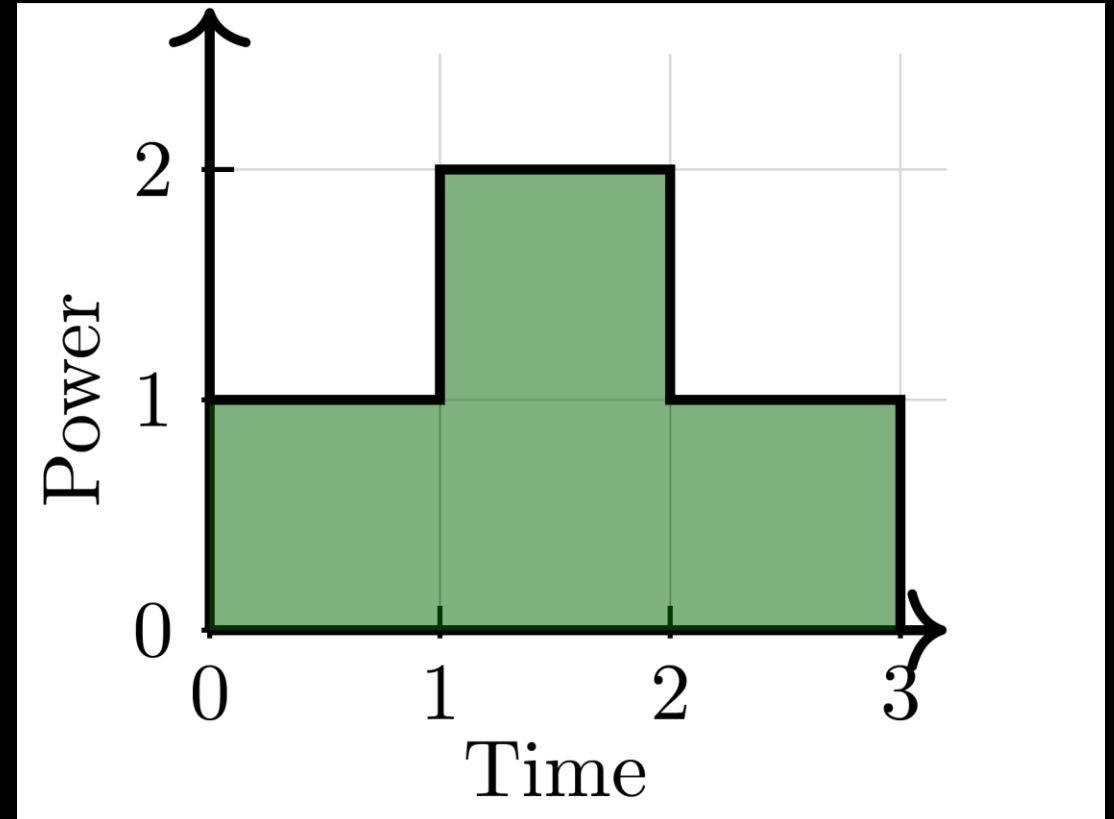
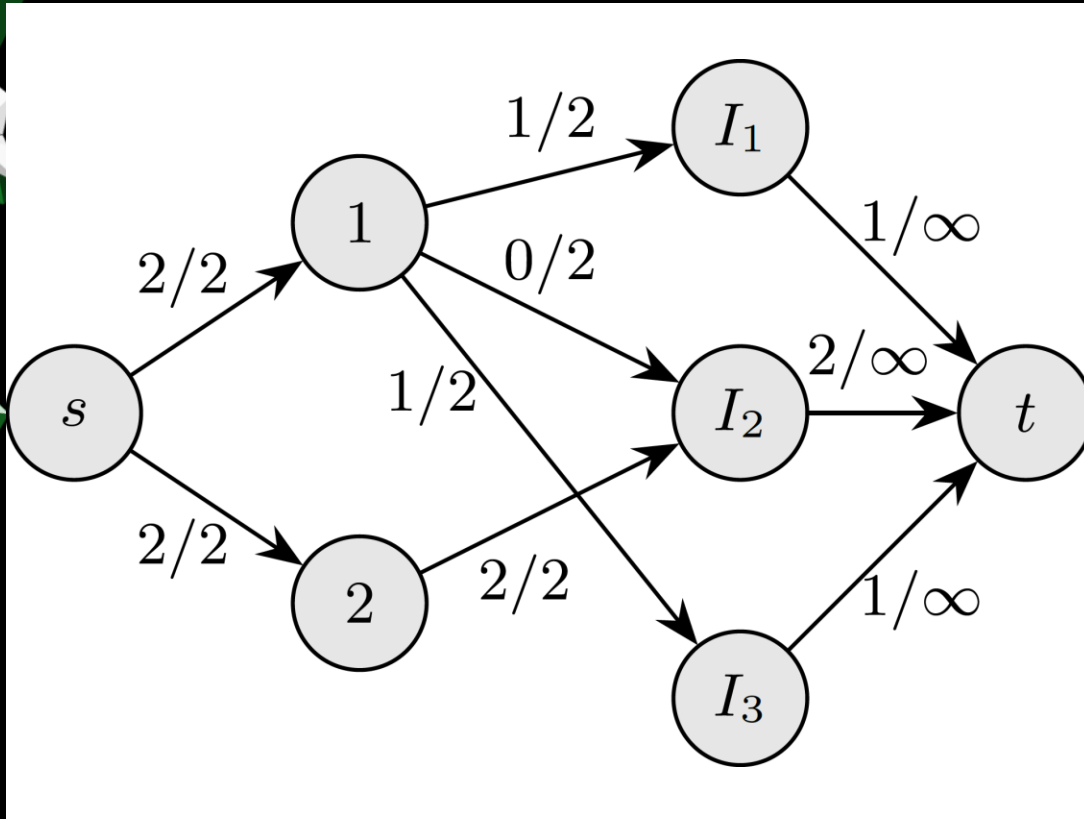
FOCS – ILLUSTRATIVE EXAMPLE

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FOCS – ILLUSTRATIVE EXAMPLE

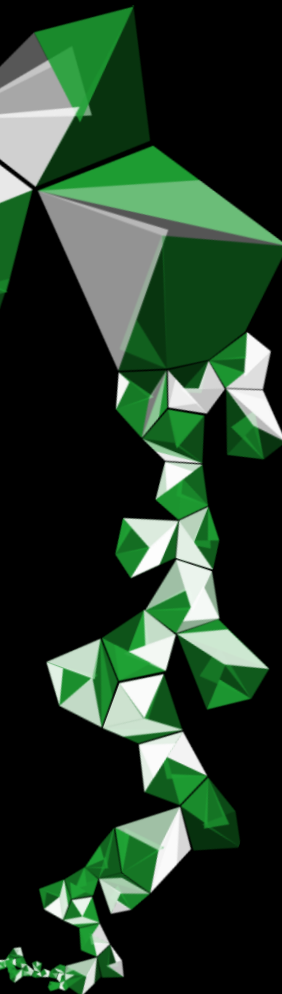
j	a_j	d_j	e_j	$p_j^{\max}$
1	0	3	2	2
2	1	2	2	2



FOCS ALGORITHM

- Flow-based Offline Charging Scheduler (FOCS)
- $O(n^2 \mu)$
- Optimal for objective functions that are:
 - Functions of the aggregated power (seperable)
 - Increasing
 - Differentiable
 - Convex

SKETCH OF OPTIMALITY PROOF

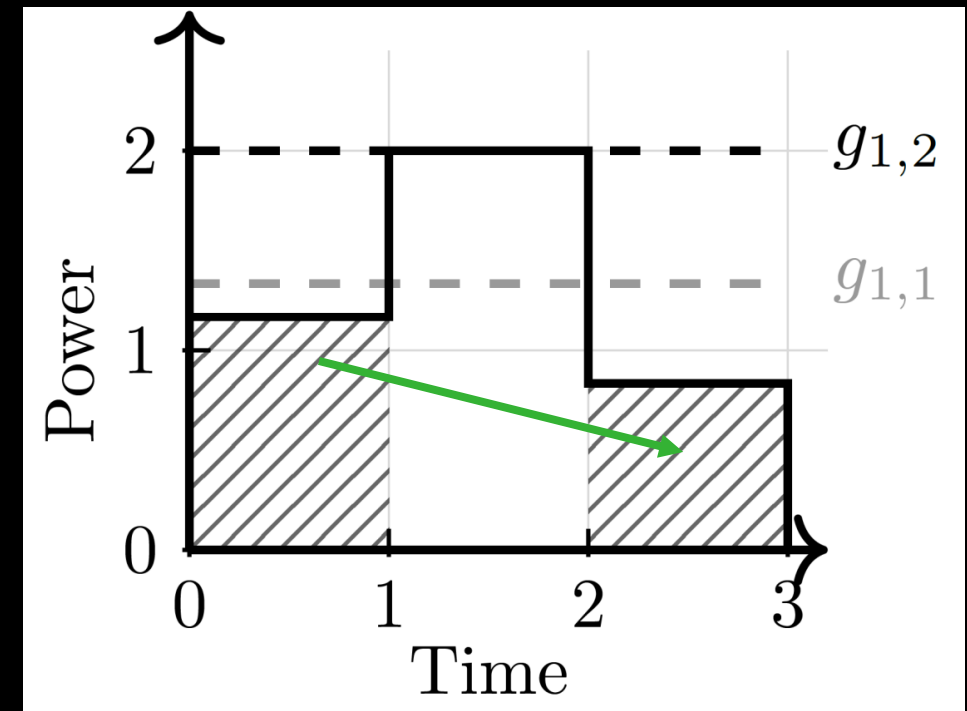
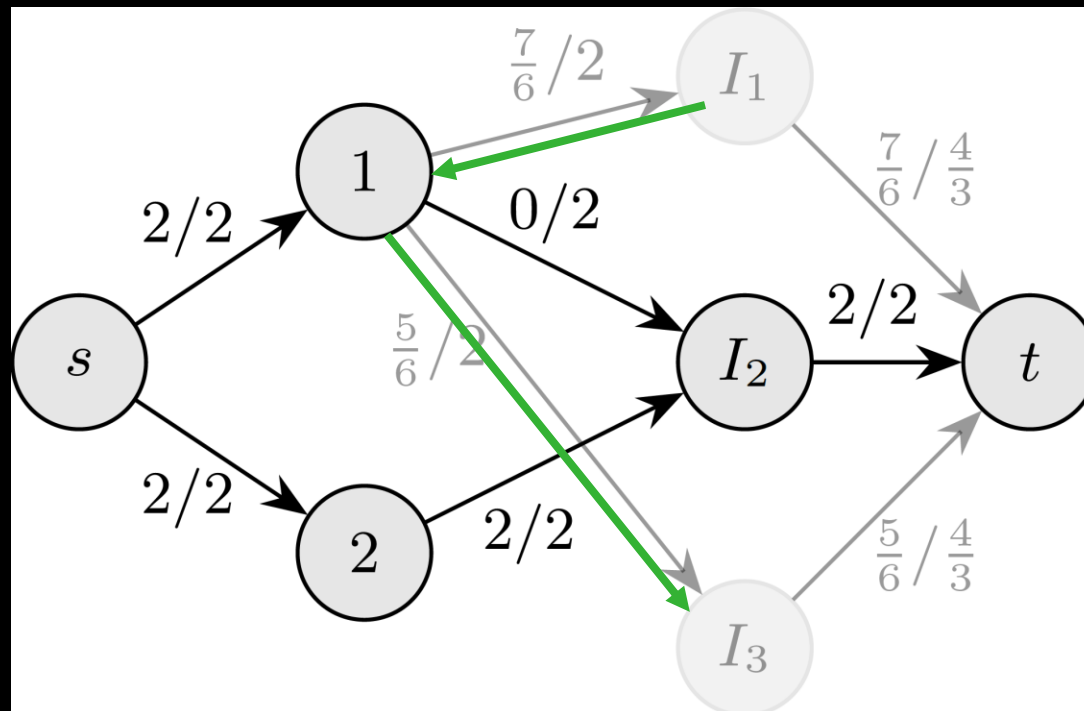


Lemma 1: The following are necessary and sufficient optimality conditions for a schedule solving speed scaling with job-specific speed limits:

- The aggregated power in all intervals where j charges but does not reach its power-limit is the same.
- The aggregated power in intervals where j could but does not charge is at least as high as where j is actually charging.
- The aggregated power in intervals where j charges at maximum power is smaller or equal than in intervals where j is charging and does not reach its limit.

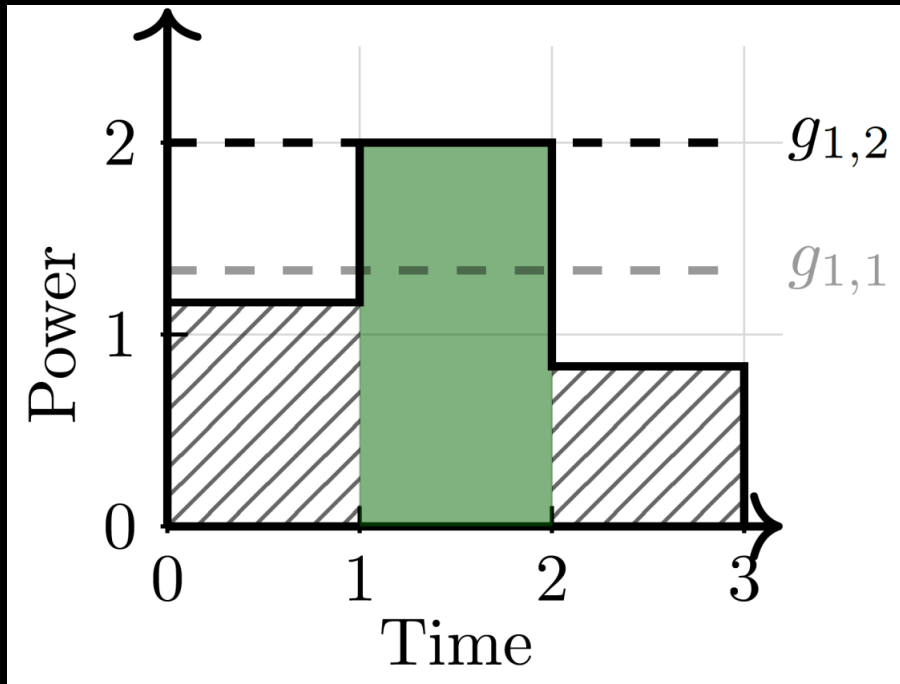
SKETCH OF OPTIMALITY PROOF

- Definition 1: Work-transferability



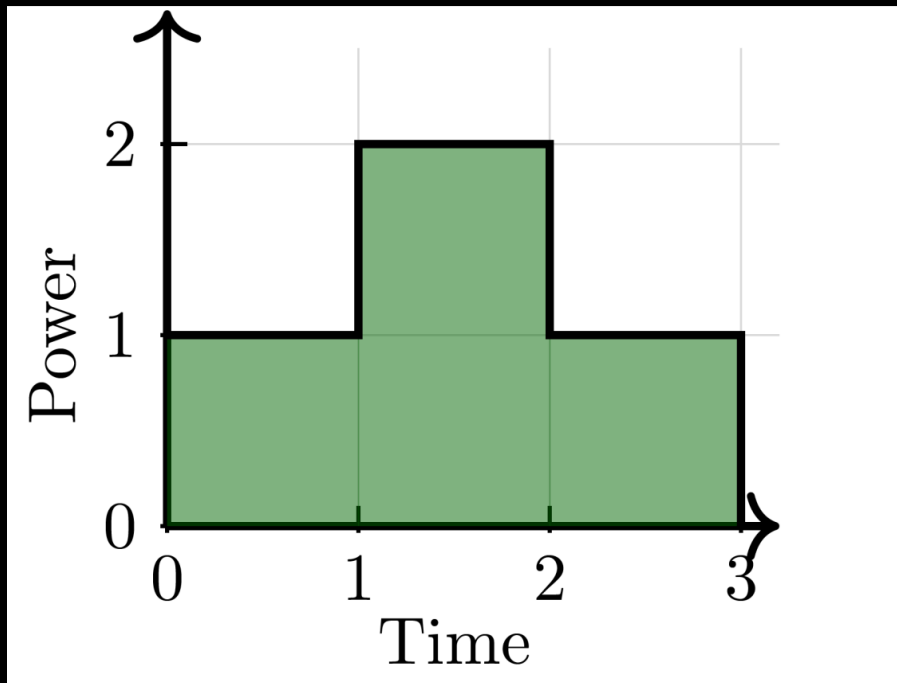
SKETCH OF OPTIMALITY PROOF

Lemma 2: Isolation of critical intervals



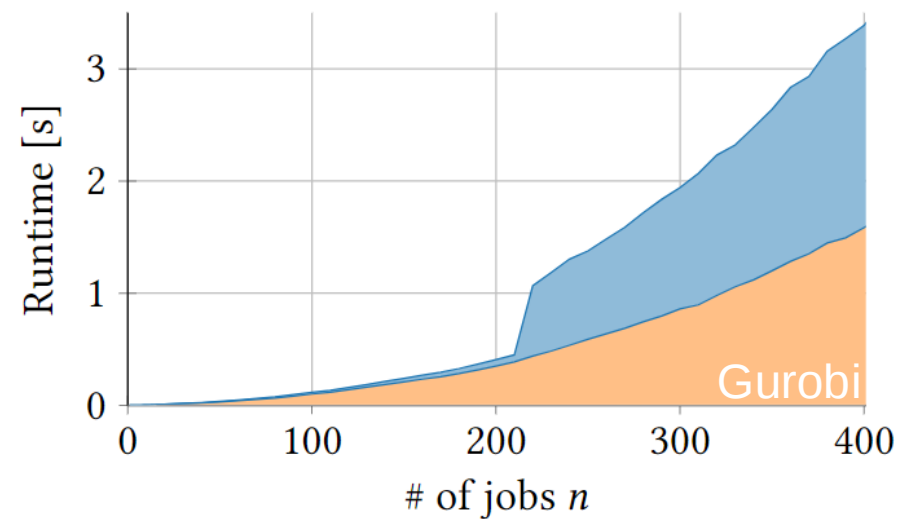
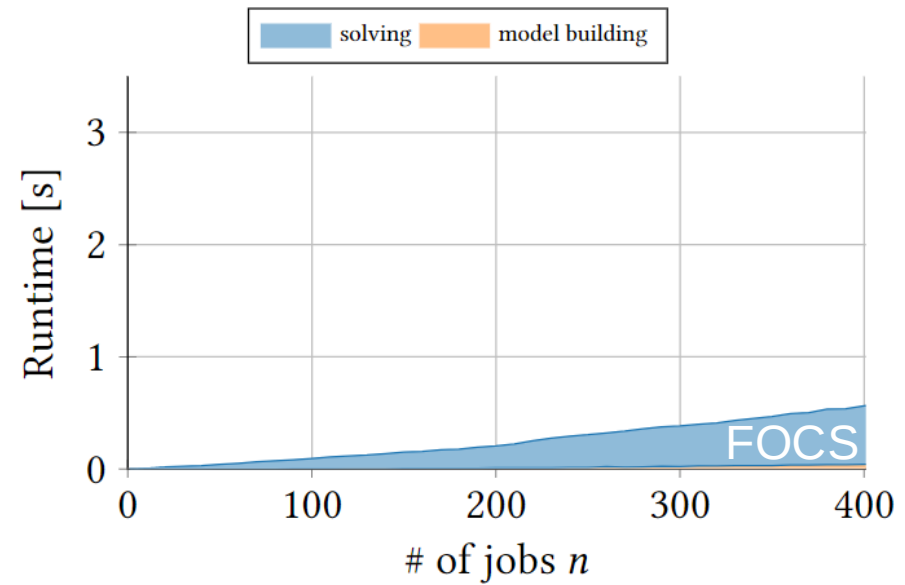
SKETCH OF OPTIMALITY PROOF

Lemma 3: Monotonicity



OUTLOOK

- Numerical experiments
- Performance guarantees of online algorithms
- Learning augmentation



CONCLUSION

- EV scheduling as speed scaling with job-specific speed limits
- FOCS as optimal algorithm

Project website:



FOCS paper: <https://arxiv.org/abs/2309.06174>



SKETCH OF OPTIMALITY PROOF

Definition 1: (Work-transferability)

For a given schedule and atomic intervals I_i and $I_{i'}$, the work-transferable relation $i \rightarrow i'$ holds if there exists a job $j \in J(i) \cap J(i')$ such that $e_{i,j} > 0$ and $e_{i',j} < e_{i',j}^{\max}$. Furthermore, let $\rightarrow$ be the transitive closure of $\rightarrow$.

SKETCH OF OPTIMALITY PROOF

Lemma 2: (Isolation of critical intervals)

If I_i is a critical interval in round r and if the round consists of multiple iterations whereby $I_{i'}$ was subcritical in one of those iterations, then there is no work-transferable relation between i and i' in the schedule corresponding to the flow at the end of round r , i.e., $i \not\Rightarrow i'$ with respect to flow f_r .

SKETCH OF OPTIMALITY PROOF

Lemma 3: (Monotonicity)

Given the schedule corresponding to output flow f of algorithm FOCS and atomic intervals I_i and $I_{i'}$, where $r(i) < r(i')$, the aggregated power in I_i is strictly larger than in $I_{i'}$:

$$p_i > p_{i'}$$

NUMERICAL RESULTS

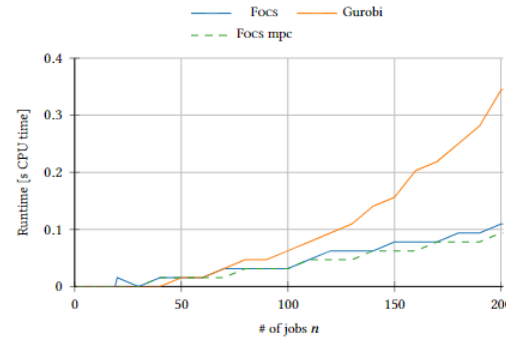
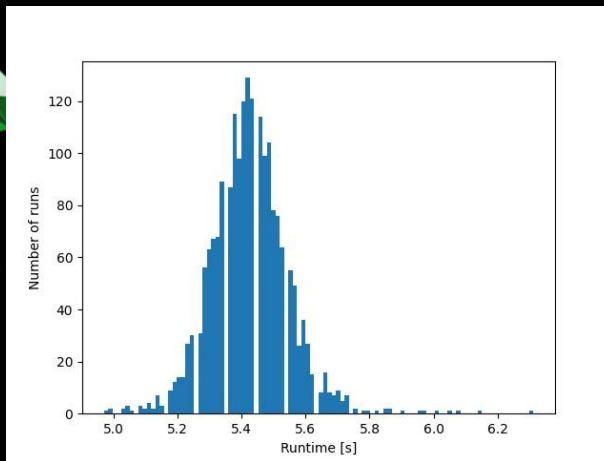
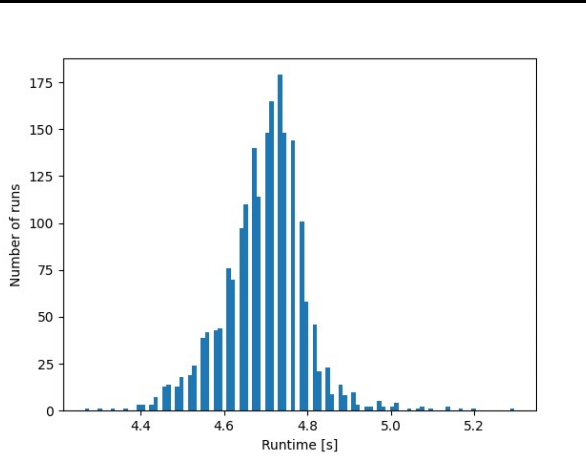


Figure 4: Runtimes for instances with realistic instance sizes, timestep = 3600s, and using shortest path embedded in Focs.

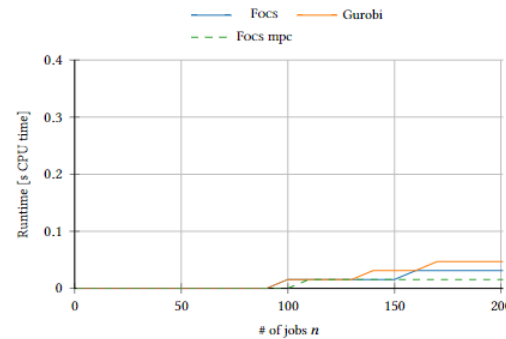


Figure 5: Runtimes for instances with realistic instance sizes, timestep = 3600s, and using shortest path embedded in Focs.

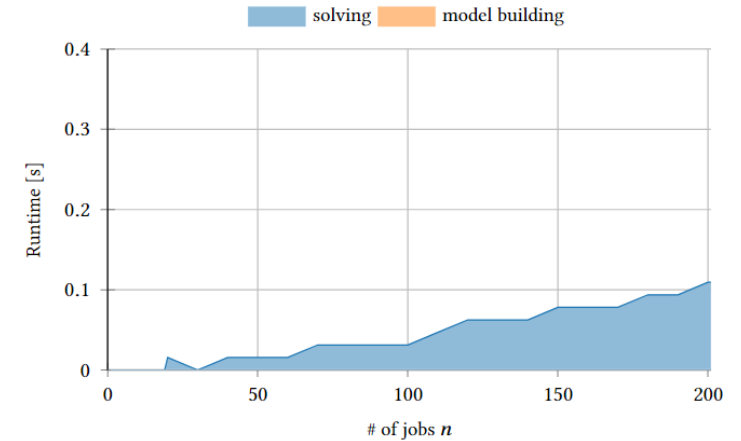


Figure 11: Runtimes of Focs. Optimized in hour-granularity.

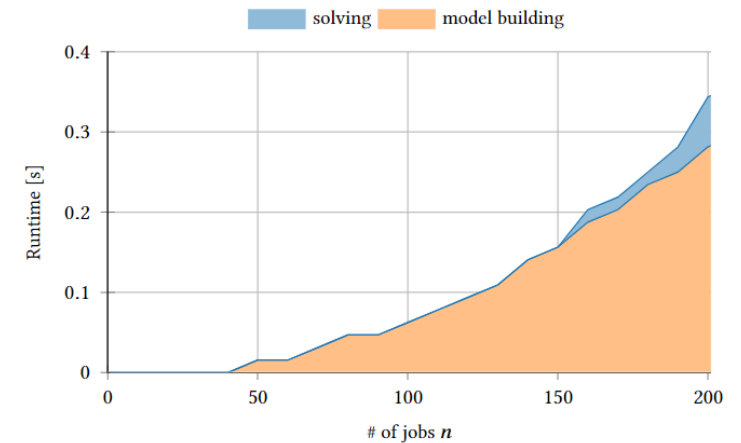


Figure 12: Runtimes of Gurobi. Optimized in hour-granularity.