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Optimal scheduling of hybrid power plants

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Hybrid power plants

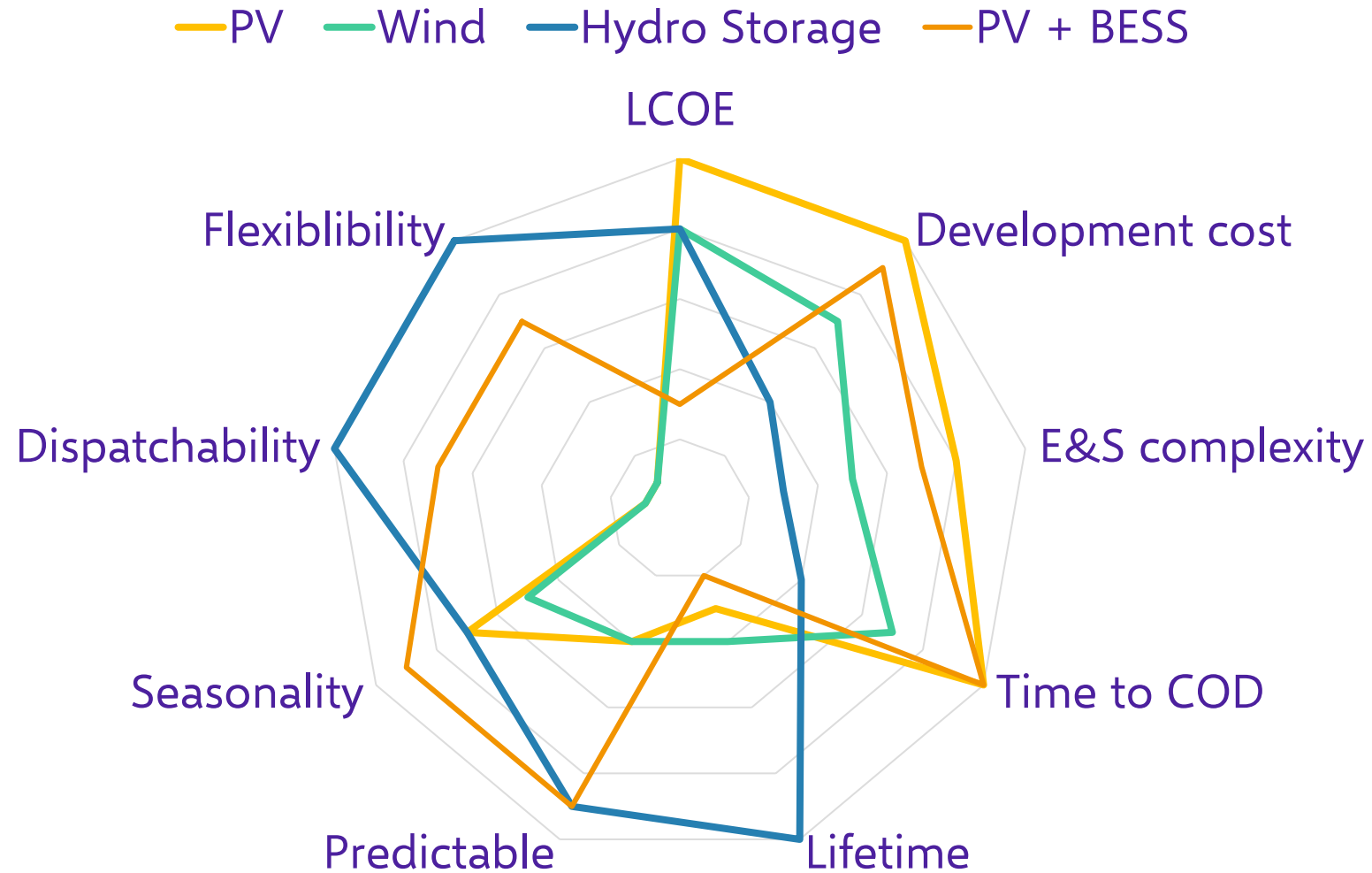
- Future energy supply will be based on renewable technologies: wind, solar and hydro...
- Solar and wind are becoming cost-competitive, but lack important capabilities in terms of storage, reliability and supplying ancillary services
- By combining the cheap energy from solar and/or wind with the regulating capabilities of hydropower, hybrid power plants have a vast potential for supplying affordable, secure and robust energy globally.



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Comparing characteristics of renewables

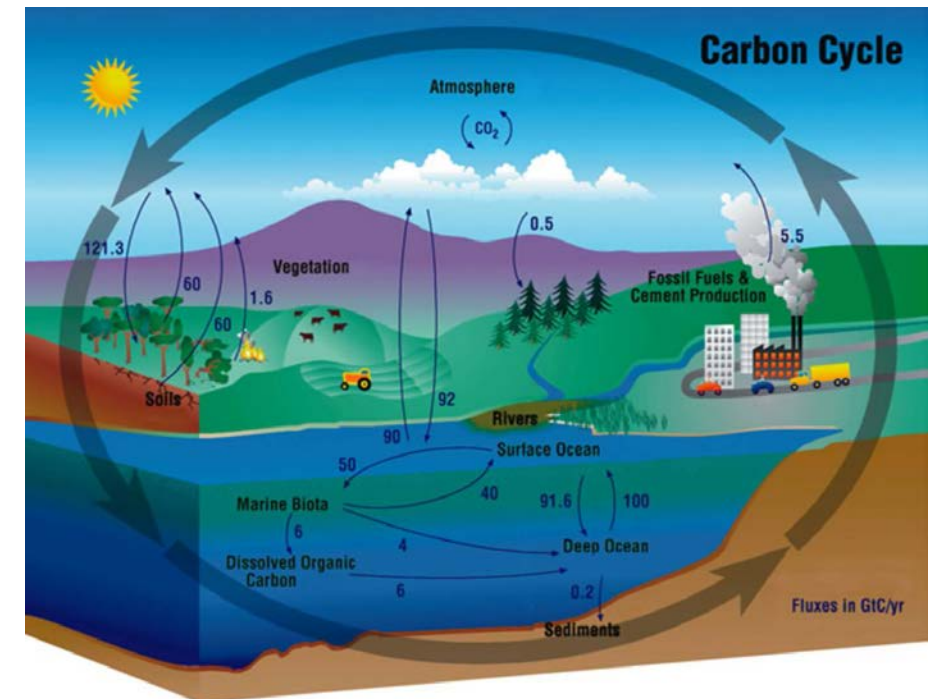
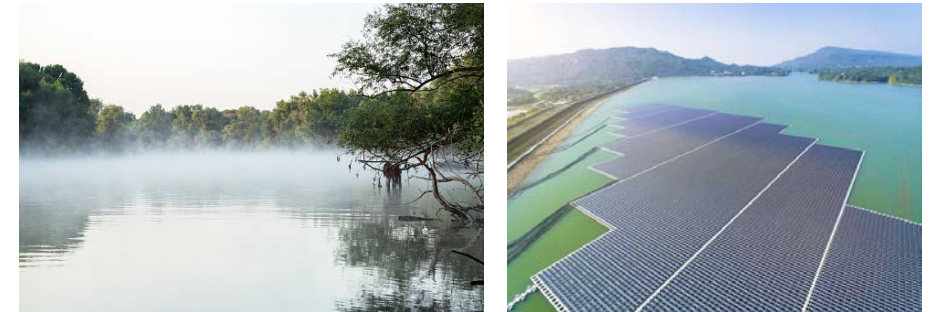




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Floating-PV and hydro hybrid power plants

- Efficient use of land areas and resources
 - Surface area of hydro reservoir is "free"?
 - Common grid connection point
 - Inflow and irradiation - correlation and seasonality?
- Floating-PV limits evaporation from hydro reservoirs
 - But how will it affect other environmental/ecosystem/biology aspects of the reservoirs and rivers?
- Constantly balancing PV will lead to more wear on hydro equipment and rapid changes in water flows
 - Battery storage may help
- Hybridization for single systems or through the market?
 - Each producer balancing their own portfolio may lead to lower liquidity in the market?
 - But very good in areas where the power market/grid is still under development





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Typical scheduling challenges

For hydro alone:

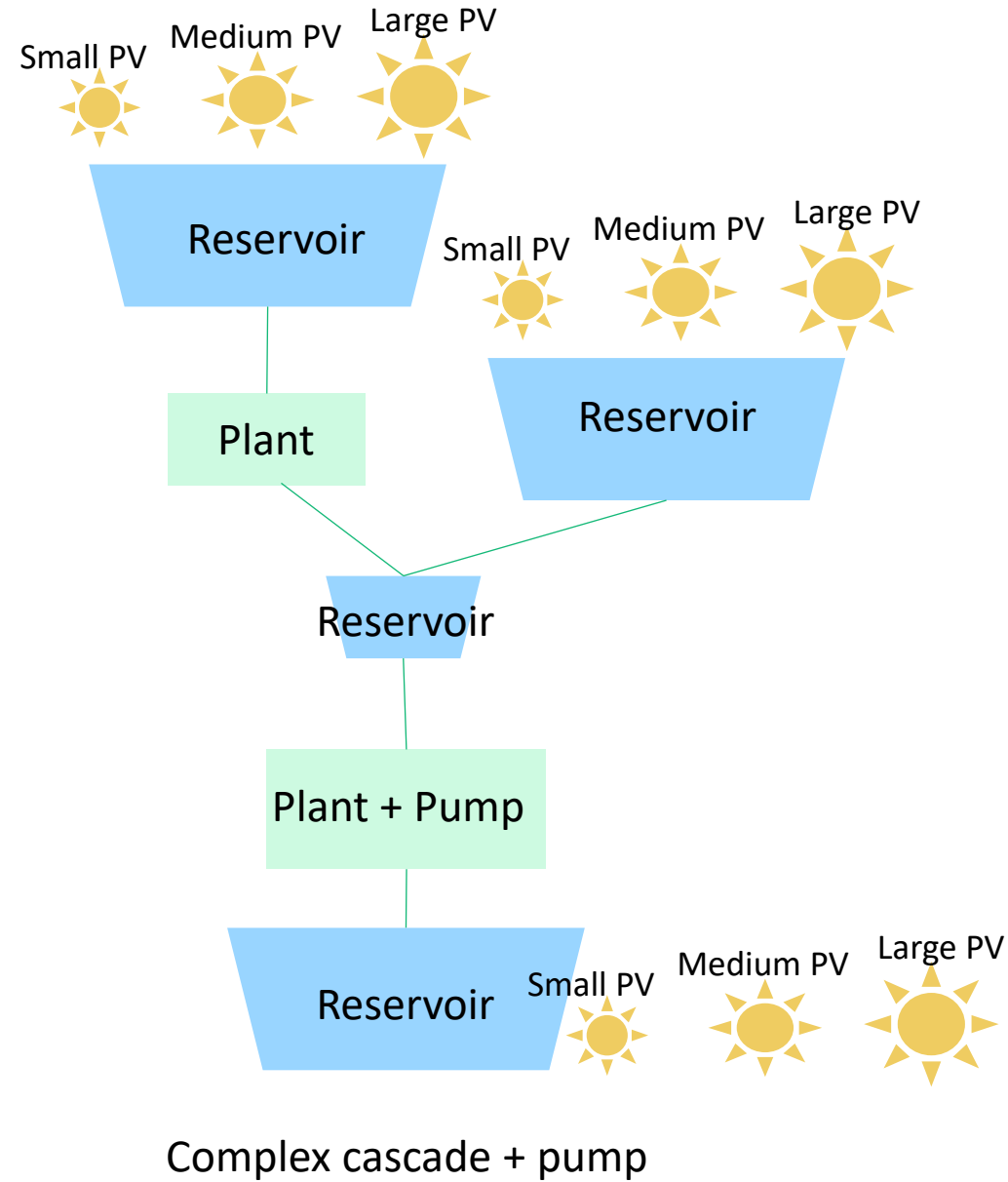
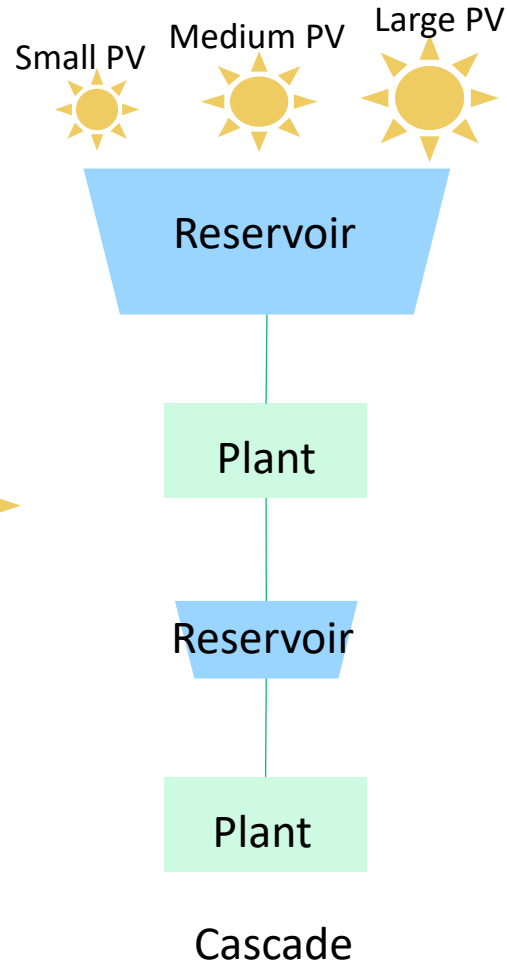
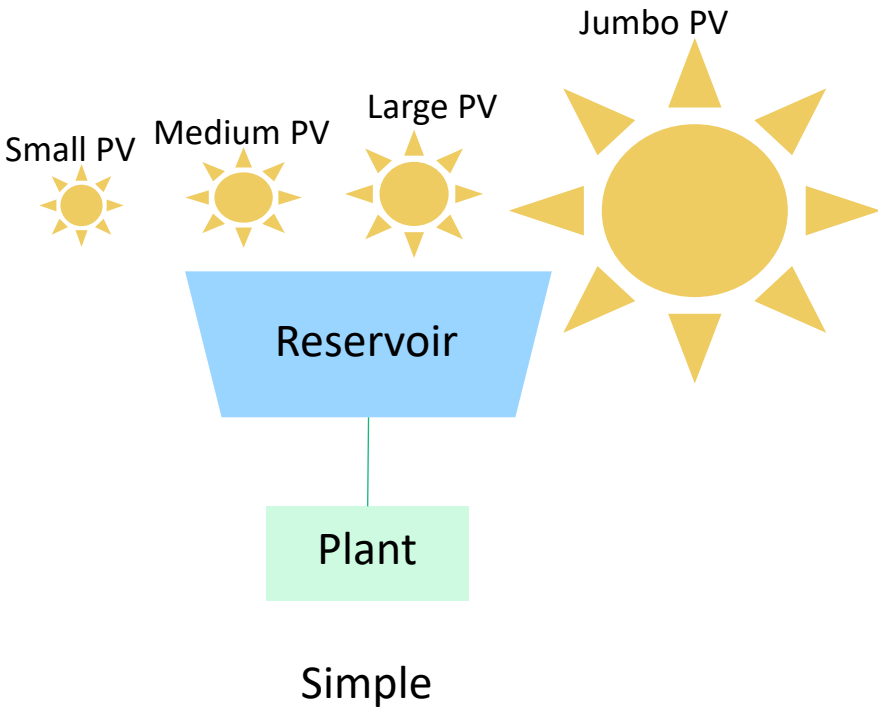
- Complex hydro systems with varying sized reservoirs in series/parallel or other complex configurations
- Complex hydraulic couplings, flow patterns in rivers and tunnels
- Wear, costs and reduced lifetime of equipment due to balancing/flexible operations
- Tight environmental constraints in cascaded systems – minimum flows, ramping, reservoir level, state-dependent constraints...
- Complex market commitments – energy, balancing, reserves...
- Uncertainty in inflow and prices, forecasting accuracy, historical data and models vs climate change

For hybrid plants:

- All of the above!
- Uncertainty for solar (both seasonal, short-term and very-short term) and also joint/correlated uncertainty for solar/inflow/prices/load
- Short-term variability of solar necessities more frequent re-optimizations, .ie. Larger potential for autonomous scheduling

Hypothesis: Large PV capacity compared to hydro capacity will amplify the problems?

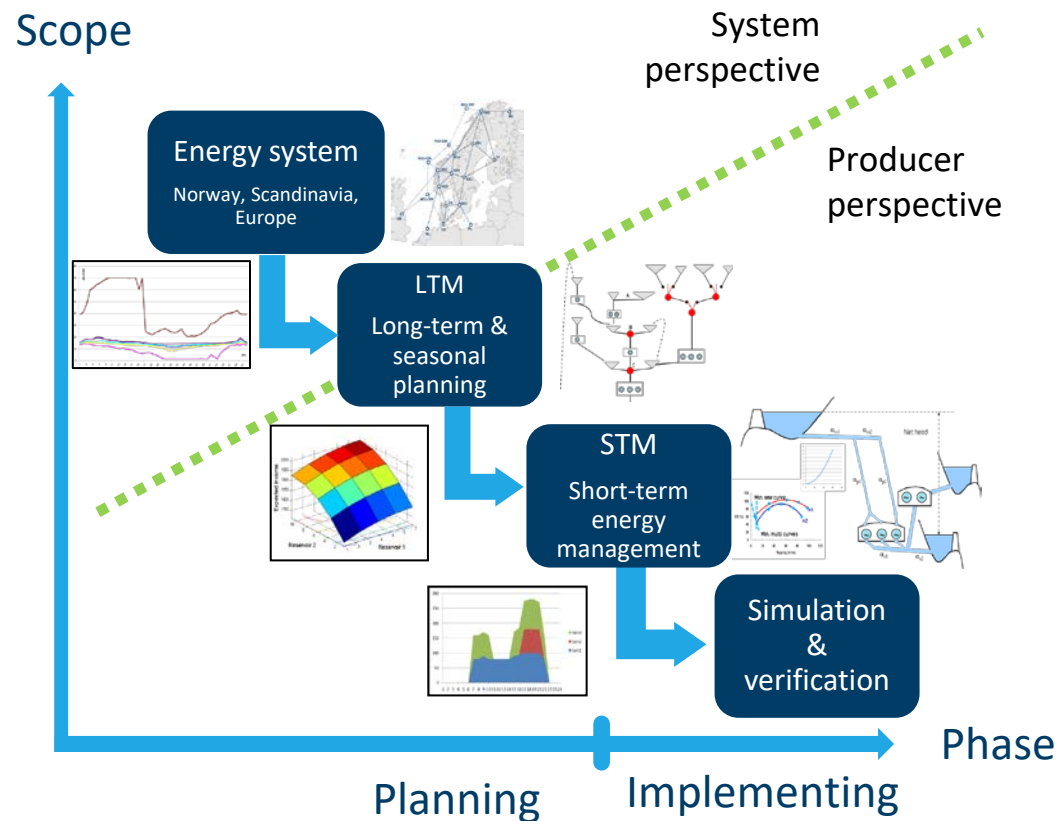
Example cases



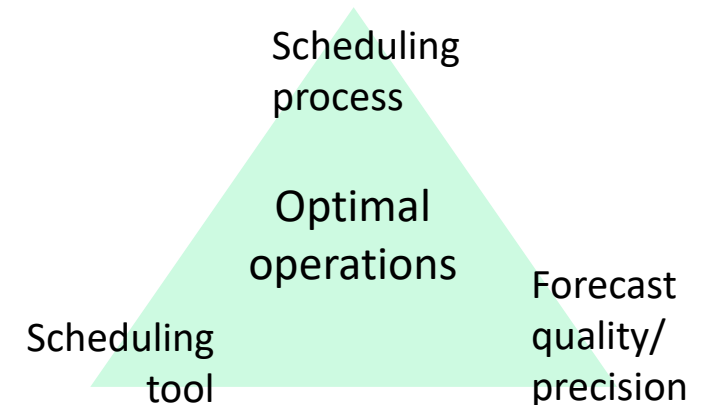


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SINTEF's hydropower models



- We have set up analysis for hybrid plants using SINTEF's hydropower models
 - STM: SHOP
 - LTM/seasonal: ProdRisk
- Modelled PV as a "solar market" where energy can be bought at zero cost
- Solar market + hydro production = cover market load

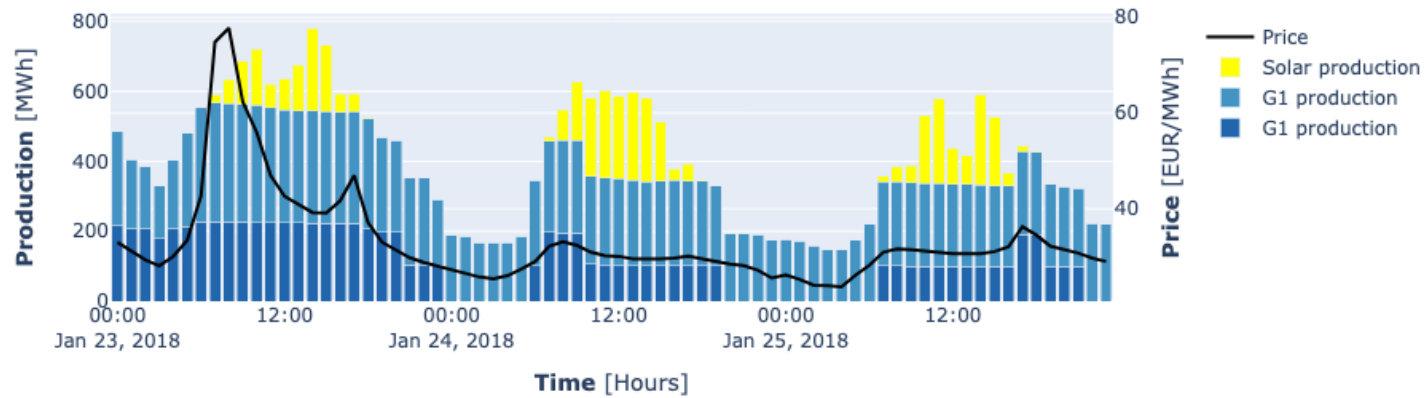




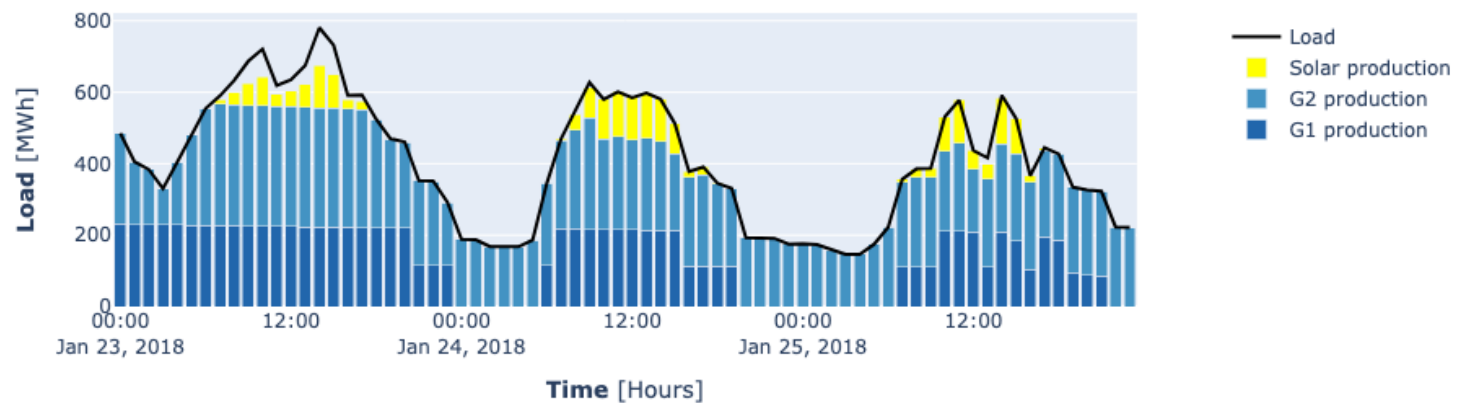
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Initial POC for short-term scheduling (SHOP)

Price vs. production



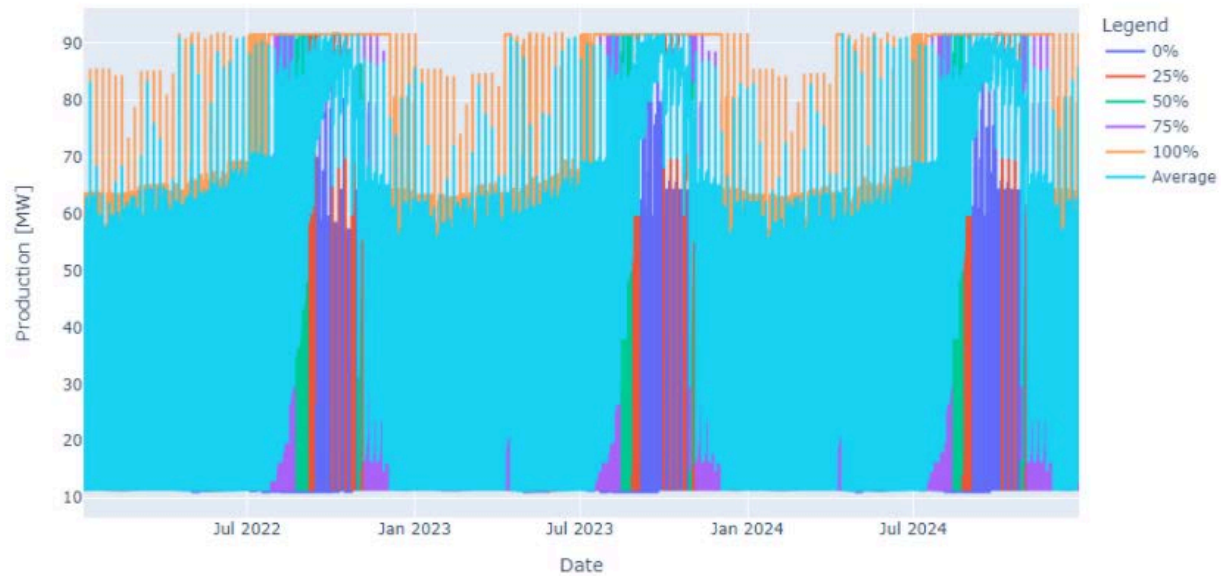
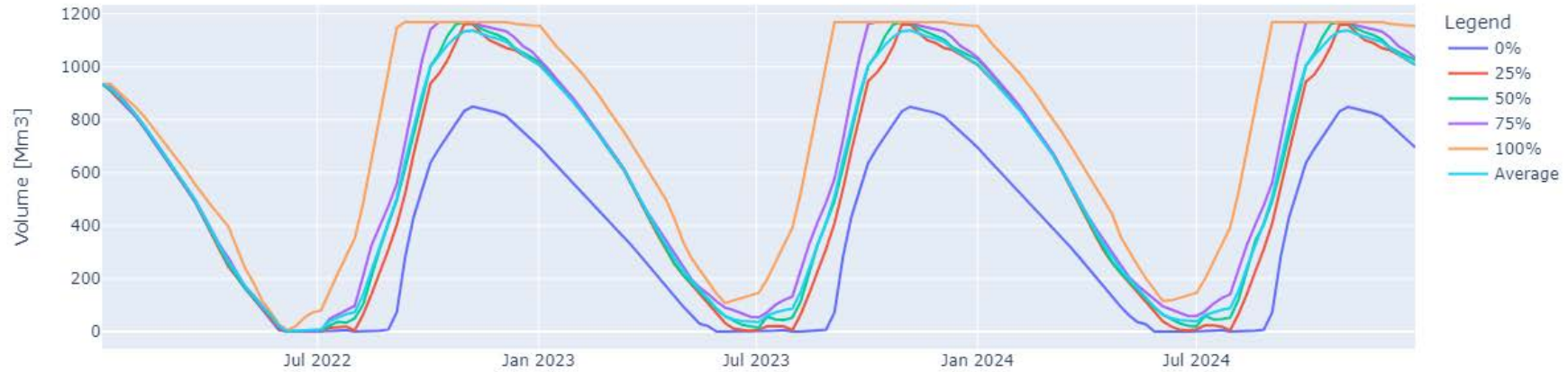
Load vs. production





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Initial POC for seasonal scheduling (ProdRisk)



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interaction	29 minutes ago
maintenance	29 minutes ago
multi_price_bid_matrix	4 minutes ago
overview	29 minutes ago
pyshop	29 minutes ago
ramping	29 minutes ago
reserve_capacity	29 minutes ago
simulation	29 minutes ago
tunnel	2 minutes ago
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yaml_standard	29 minutes ago
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Running SHOP

Once the model is fully defined, we can prepare for a call to the optimizer. It is possible to define certain criteria depending on the solver used, if not, default values will be effective.

In order to find an optimal solution, it is normal to run SHOP with multiple iterations, both full and incremental.

```
[32]: # Setting a full flag, telling SHOP the upcoming iterations should be full
shop.set_code(['full'], [])

# Starting SHOP and running five (full) iterations
shop.start_sim([], ['5'])

# Setting an incremental flag, telling SHOP the next iterations should be incremental
shop.set_code(['incremental'], [])

# Running three more (incremental) iterations
shop.start_sim([], ['3'])
```

[32]: True

Results

After the optimization has completed, we can review the result from SHOP by plotting the graphs we want.

...

Price vs. production

```
shop.model.plant.plant2.connect_to(shop.model.reservoir.rsv3)
```

The resulting topology is shown below. Note that the creek intake is modelled as a small reservoir where the volume represents the cross cut from the intake to the tunnel.

```
[6]: shop = new_model()
add_reservoirs(shop)
add_tunnels(shop)
add_plants(shop)
connect_objects(shop)
shop.model.build_connection_tree()
```

```
[6]:
```

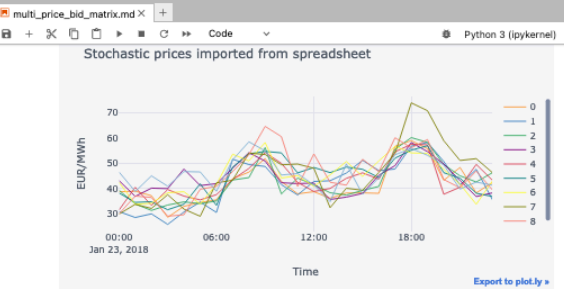
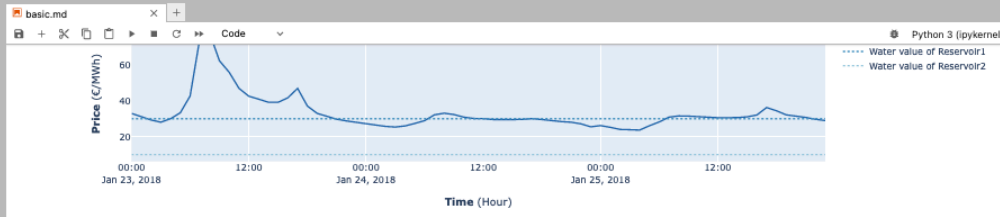
The model is initialized with historical spot prices from NO1 and higher water value for Reservoir1 than the others.

```
[7]: def init_model(shop):
time_res = shop.get_time_resolution()
starttime = time_res['starttime']
endtime = time_res['endtime']

price = pd.Series(
[189.13, 187.4, 184.28, 179.79, 178.55, 184.04, 184.96, 185.19, 184.4]
index=[starttime+pd.Timedelta(hours=t) for t in range(24*4)]
)
```

```
top - 09:49:24 up 1 day, 1:18, 0 users, load average: 0.89, 1.57, 1.15
Tasks: 14 total, 1 running, 13 sleeping, 0 stopped, 0 zombie
%Cpu(s): 5.7 us, 2.9 sy, 0.0 ni, 91.1 id, 0.2 wa, 0.0 hi, 0.2 si, 0.0 st
MiB Mem: 7957.5 total, 174.3 free, 3985.2 used, 3877.9 buff/cache
MiB Swap: 0.0 total, 0.0 free, 0.0 used, 3749.3 avail Mem
```

PID	USER	PR	NI	VRT	RES	SHR	S	%CPU	MEM	TIME	COMMAND
7	jovyan	20	0	987464	213664	22244	S	1.0	2.6	0:39.34	ipyterhub-sing
548	jovyan	20	0	10432	4096	3336	R	0.3	0.1	0:00.19	top



Creating scenarios

Now it is time to create scenarios that we can populate with the imported prices. We make sure to create just as many scenarios as we have prices imported.

```
[12]: # Generate as many scenarios as prices
n_scenarios = n_prices
for i in range(1, n_scenarios+1):
scenario_name = S+'str'+i
# The first scenario always exists in SHOP and should not be added again
if i==1:
scenario = shop.model.scenario.add_object(scenario_name)
else:
scenario = shop.model.scenarios[scenario_name]
scenario.scenario_id.set(i)

# Set each scenario equally probable
scenario.probability.set(1.0/n_scenarios)

# Branch immediately, i.e. at 'starttime'
scenario.common_scenario.set(pd.Series([i], index=[timeres['starttime']]))

# Optionally set branching to start after given number of hours (all scenarios)
#scen.common_scenario.set(pd.Series([2, 1], index=[timeres['starttime'+i, branch:
```

Creating the new price array from stochastic and deterministic prices

Since we have chosen to only consider the first 24 hours as stochastic when it comes to the price, but have longer total time horizon, we need to combine the stochastic and deterministic prices into a joint price dataframe. We have already defined the start and end time for the stochastic price, but need to make sure that we also define when the deterministic price should be valid and thus overlap each other.

```
[13]: # Use first (and only) market as index for setting stochastic data and getting resu:
```

```
shop.messages.log
```

```
1 INFORMATION: 1032
2 14.4.2.1 Cplex 20.1.0 Gurobi 7.5 OSI/CBC 2.9 2022-09-05
3
4 INFORMATION: 1047
5 Current time: Tue Sep 6 09:40:12 2022
6
7 DIAGNOSIS WARNING: 3226
8 Reservoir Reservoir1:
9 Start of Head/Flow description = 906.000000, should be 905.000000
10 DIAGNOSIS WARNING: 3285
```



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More robust and flexible model coupling between SHOP and ProdRisk

[1]:

```
from pyshop import ShopSession
```

[2]:

```
shop = ShopSession()
```

[]: shop.model.

- generator
- global_settings
- inflow_series
- interlock_constraint
- junction
- junction_gate
- lp_model
- market
- needle_comb_reserve_capability



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More robust and flexible model coupling between SHOP and Prodrisk

```
[1]: from pyprodrisk import ProdriskSession  
from pyshop import ShopSession
```

```
[2]: prodrisk = ProdriskSession()  
shop = ShopSession()
```

```
[ ]: shop.model.
```

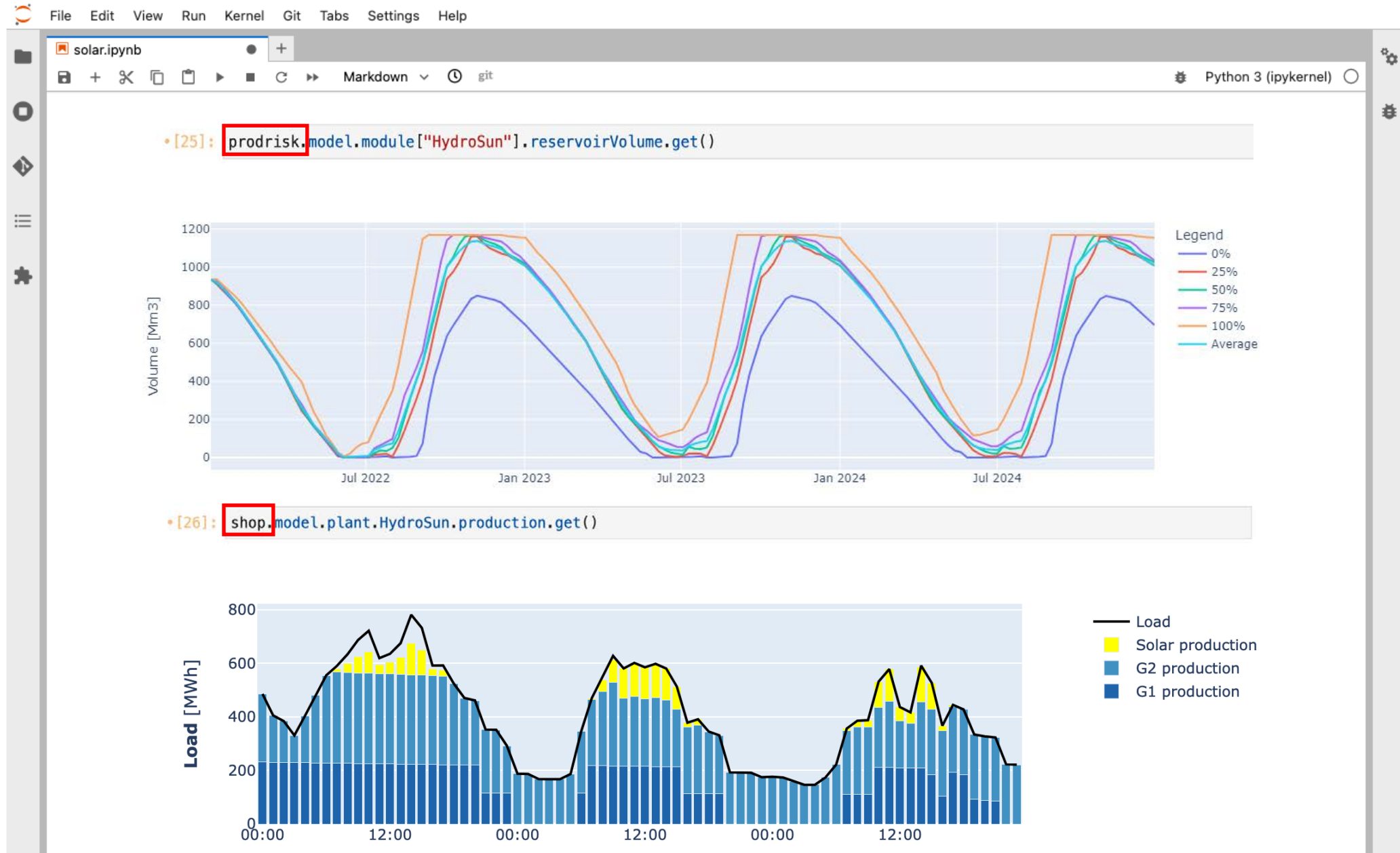
- generator
- global_settings
- inflow_series
- interlock_constraint
- junction
- junction_gate
- lp_model
- market
- needle_comb_reserve_capability

```
[ ]: prodrisk.model.
```

- contract
- effectProfile
- inflowSeries
- loadProfile
- marketStep
- module
- penaltyStep
- pricePeriod
- prodrisk_optimize
- prodriskWork



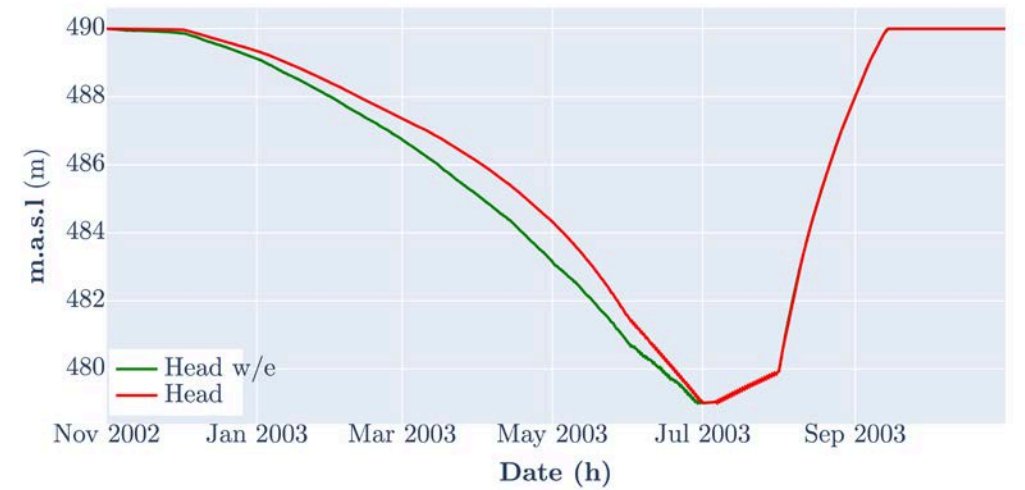
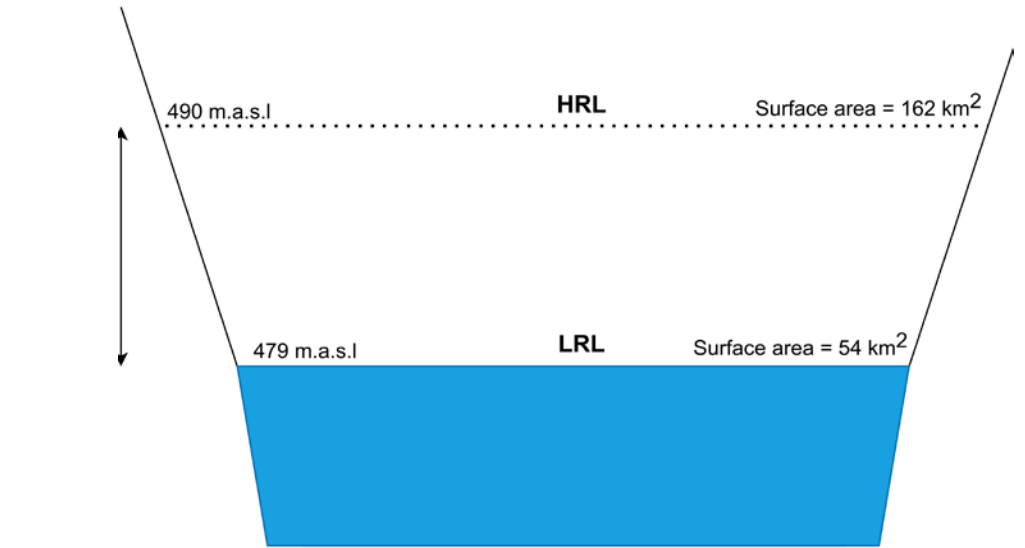
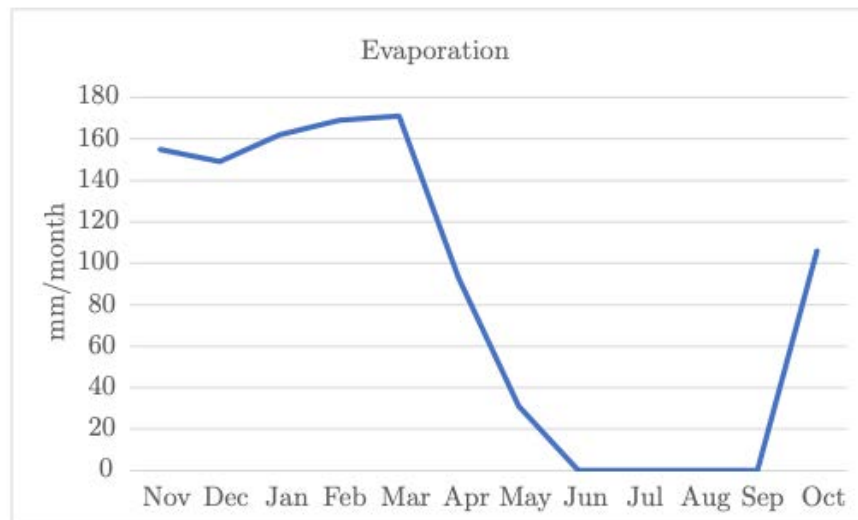
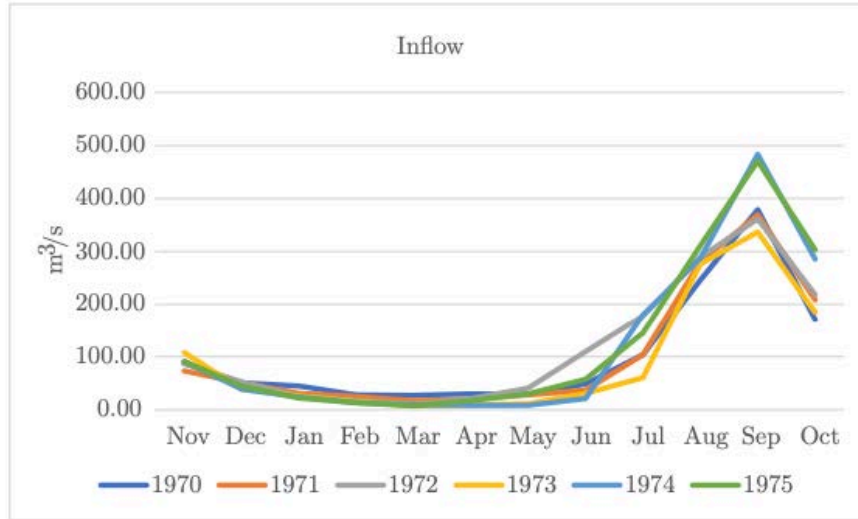
- Same (running) kernel
- Same environment
- Same user interface
- Easier debugging
- Better data handling
- More flexible
- More robust





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Evaporation

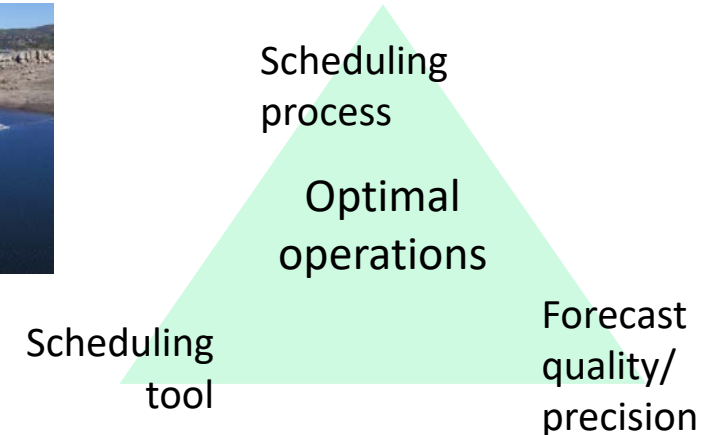




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Future work

- Project period: 2021-2024
- Need for scheduling models to be updated more frequently in order to re-plan closer to real-time
- The optimal hydro-PV hybrid scheduling system will realize the complementation of energy generation over seasons, days, hours, and seconds
- Establish a new reservoir trajectory curve that includes inflow, power generation as well as solar generation effect
- PhD + MSc students also involved





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Technology for a better society

This presentation is based on work by
Jiehong Kong, Hans Ivar Skjelbred, Hans Olaf Hågenvik, Benjamin Trondsen, Bjørnar Fjelldal and Ellen Krohn Aasgård