

Representation of uncertainty in market models for renewable power systems

Mari Haugen^{1,2}, Hossein Farahmand¹, Stefan Jaehnert², Stein-Erik Fleten³

¹Department of Electric Power Engineering, NTNU

²Energy Systems, SINTEF Energy Research

³Department of Industrial Economics and Technology Management, NTNU

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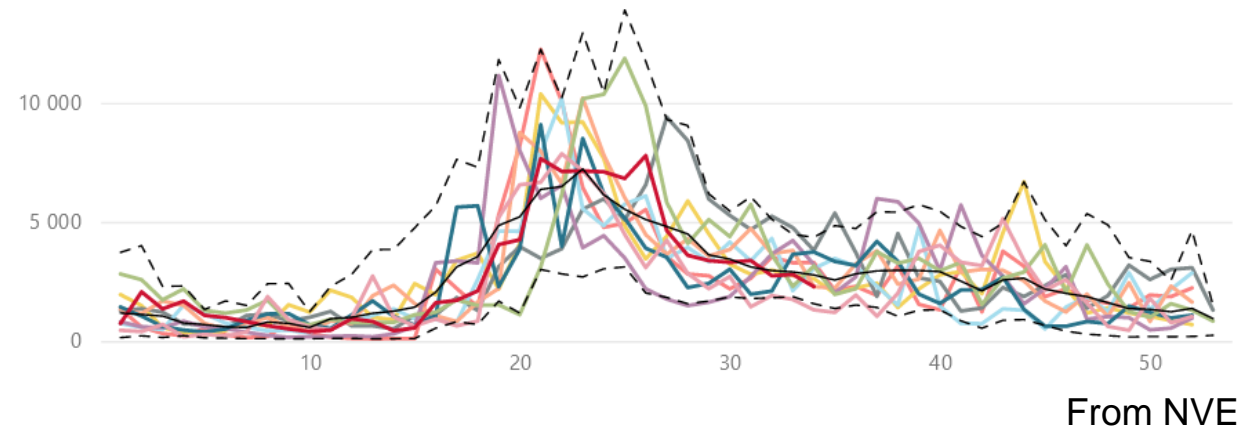


Motivation

- Operational planning of hydropower



Nyttbart tilsig per uke [GWh]



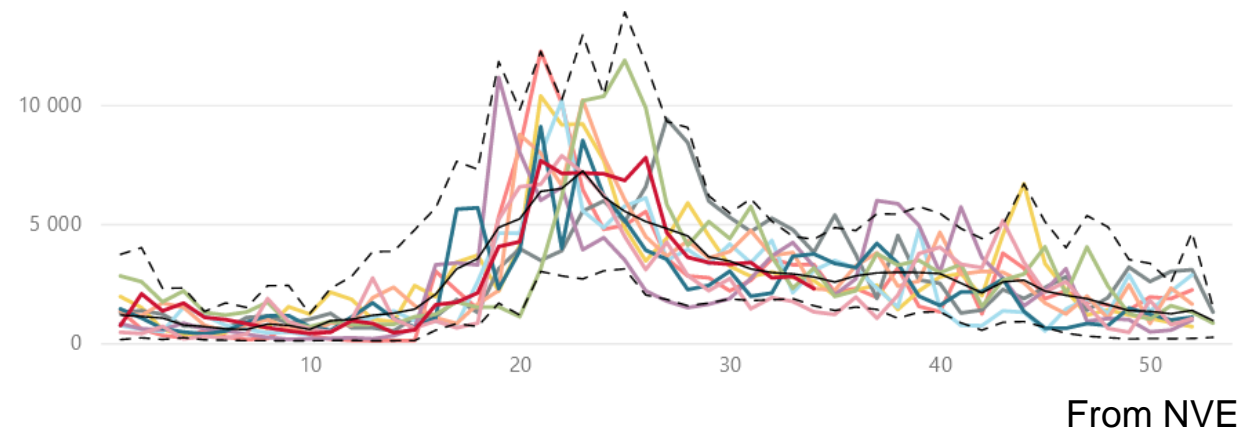


Motivation

- Operational planning of hydropower
- The future is uncertain!



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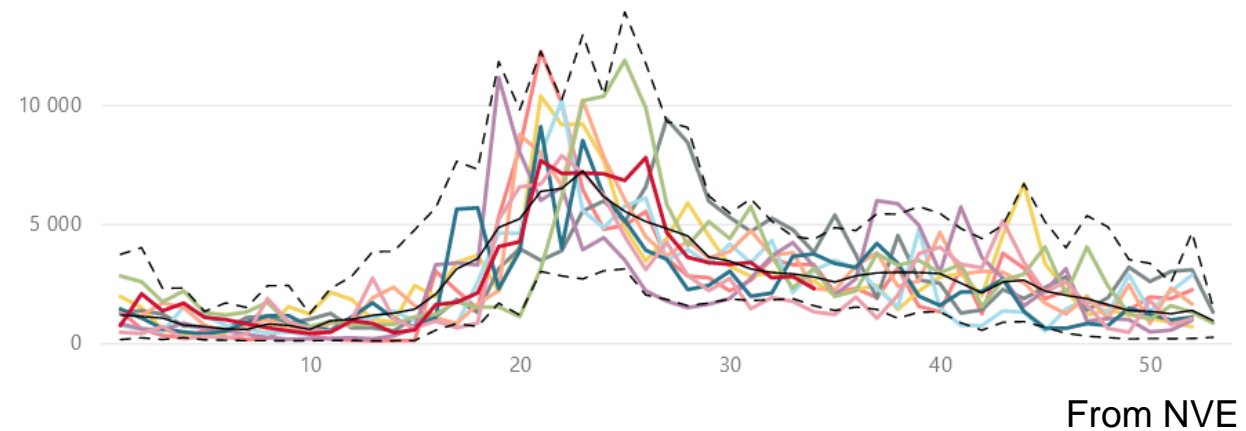


Motivation

- Operational planning of hydropower
- The future is uncertain!
- Increased need for decision support tools and increased levels of uncertainty
 - VRE integration
 - Market liberalization
 - Climate change
 - Changing regulations
 - Technology development
- Uncertainty is often overlooked or simplified in models



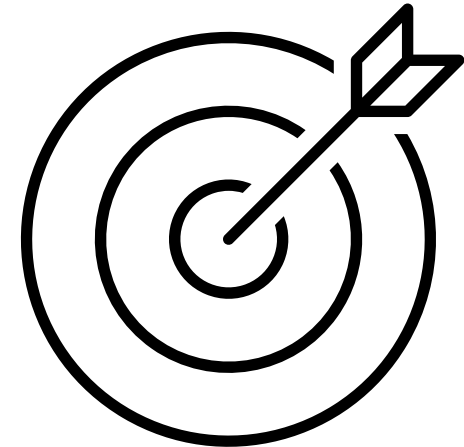
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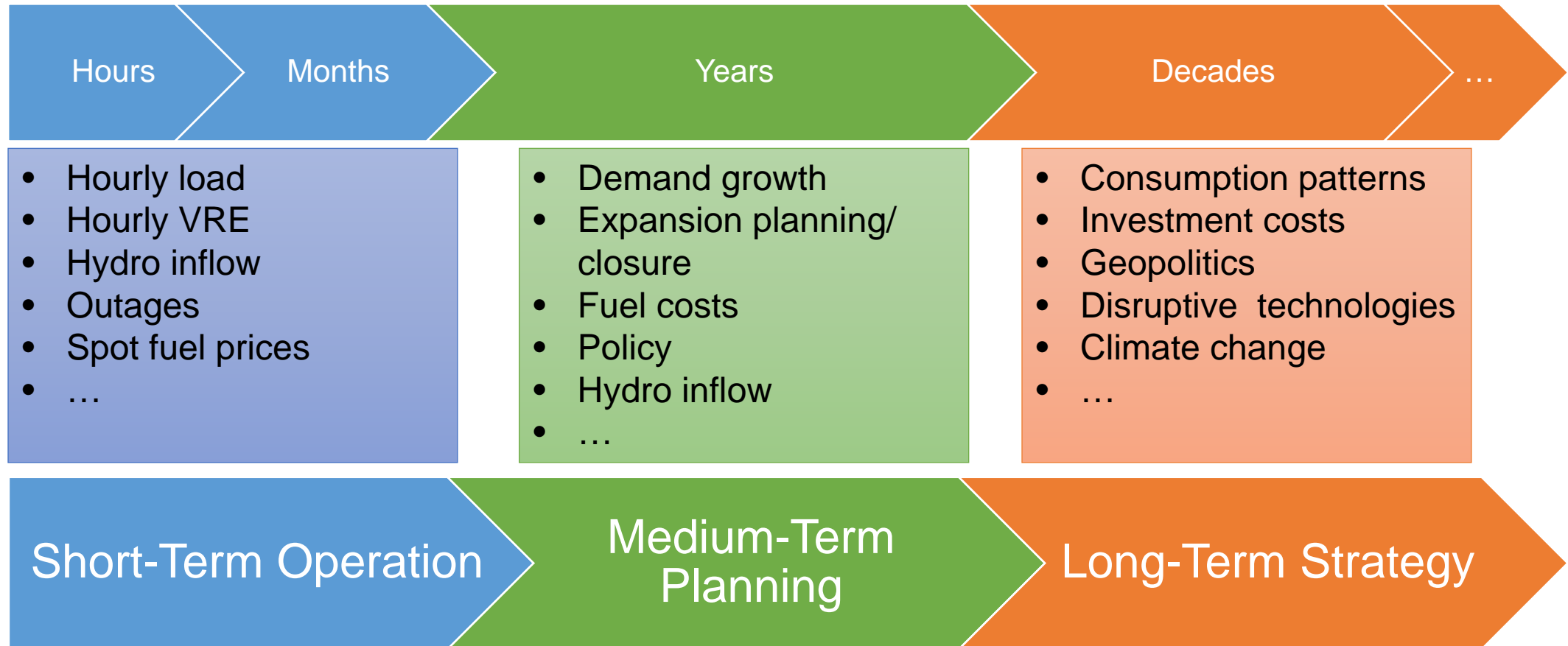
Goal of the review

- Review the representation of uncertainty in applied power market models
 - Types
 - Timescales
 - Methods
- Overview of methods:
 - Stochastic optimization
 - Scenario generation and reduction
- Identifying critical uncertainties and suitable methods





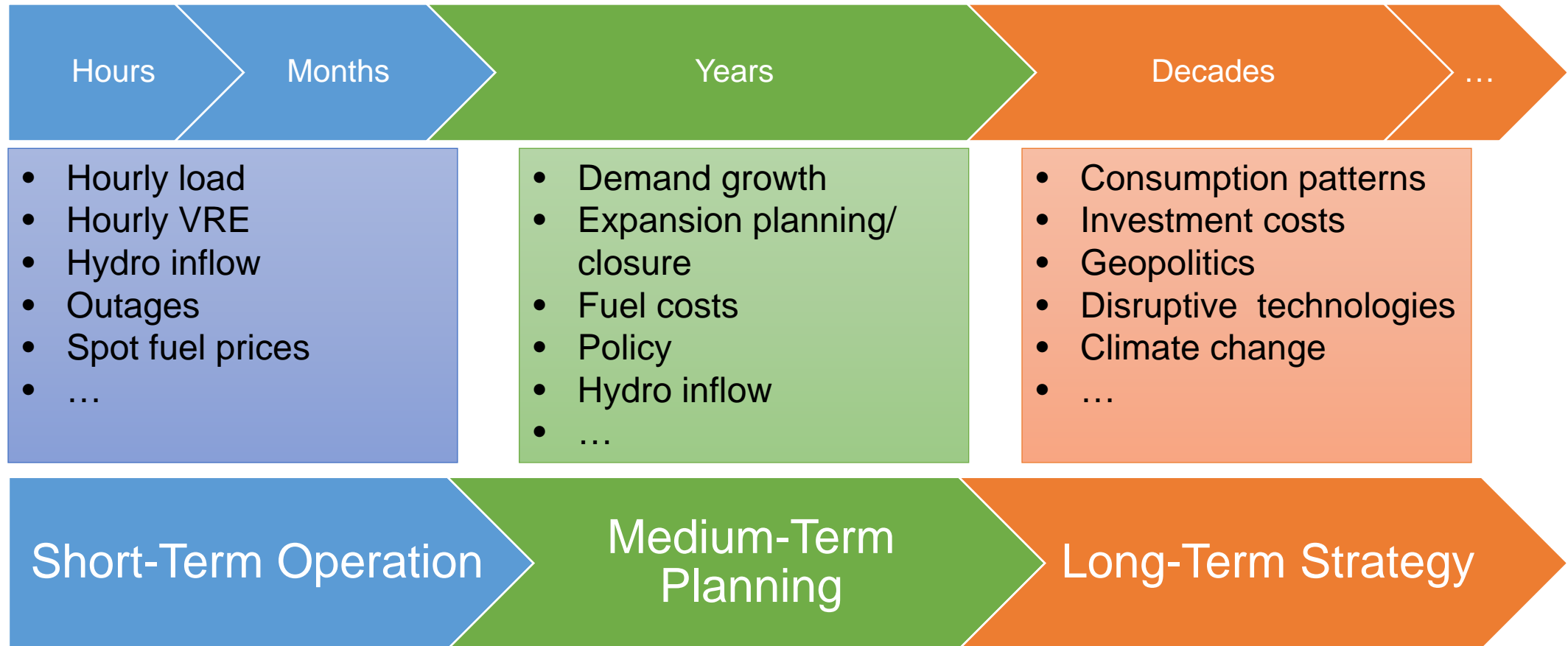
Timescale of uncertainty



Based on Velasquez et al. (2016)



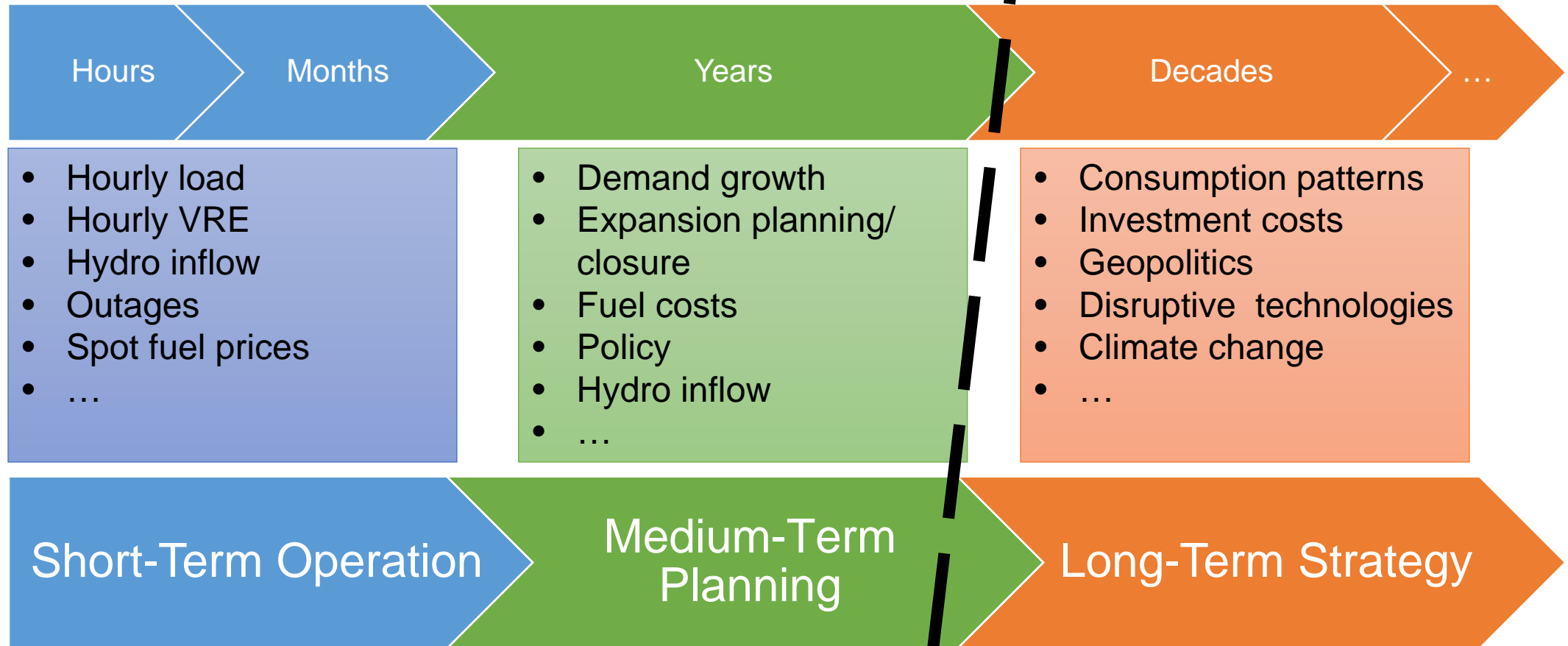
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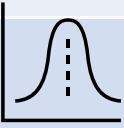
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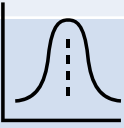
Structure of uncertainty

Uncertainty			
	«Known»	«Unknown»	«Unknowable»
Examples	<ul style="list-style-type: none"> • Demand • Weather-related power production 	<ul style="list-style-type: none"> • Outages/availability • Fuel prices/shortage • Extreme weather 	<ul style="list-style-type: none"> • Black swans • Terror and war • Natural disasters
Appropriate representation	<ul style="list-style-type: none"> • Probabilistic representation  • Stochastic process 	<ul style="list-style-type: none"> • Bounded sets/Intervals • Fuzzy sets 	<ul style="list-style-type: none"> • Scenarios
Suitable method	<ul style="list-style-type: none"> • Stochastic programming 	<ul style="list-style-type: none"> • Robust optimization • Fuzzy programming 	??

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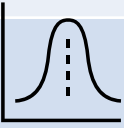
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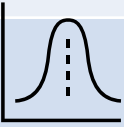
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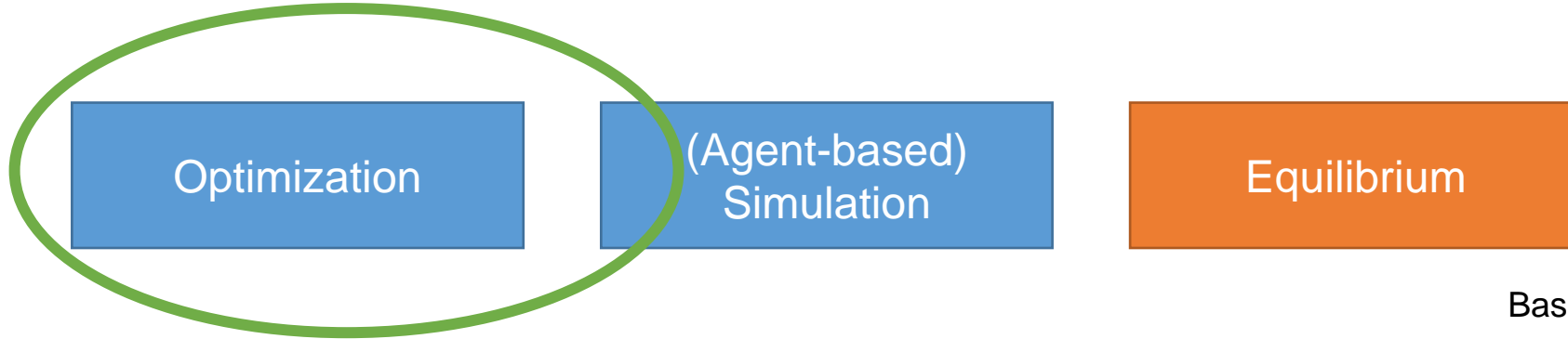
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Based on Velasquez et al. (2016)



Power market models



Based on Ventosa et al. (2005)



Power market models



Based on Ventosa et al. (2005)

Optimization

- Deterministic
- Multi-deterministic
 - Scenario analysis
 - Monte Carlo
- Stochastic
 - Stochastic programming
 - Chance constrained programming
 - Robust programming
 - Distributionally robust optimization
 - Interval programming
 - Fuzzy programming

The 'Stochastic' category and its sub-items are circled in green. Orange lines connect the dots to the right of the list items to the corresponding boxes in the main diagram above.



Power market models



Based on Ventosa et al. (2005)

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Fundamental models

- Physical description
- Explaining complex relationships
- Suitable for analysing future systems



Power market models



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- Deterministic
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Fundamental models

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Applications

- Forecast power prices
- Unit commitment and dispatch
- Operational decisions (scheduling)
- Hydrothermal coordination
- Power market analysis
- Analyse market design
- Expansion planning
-

Name	Reference	Time horizon	Problem solved	Uncertainty					Problem formulation	Solution method
				Demand	Hydro Inflow	VRES	Fuel prices	Outages		
EMPS	Wolfgang et al (2009)	Long	Hydrothermal coordination	X	X	X	(X)		multi-stage stochastic linear programming problem	SDP
FanSi	Helset et al. (2007)	Long	Hydrothermal coordination	X	X	X	(X)		two-stage stochastic linear programming problem	Benders decomposition and rolling horizon
NEWAVE	Maceria et al. (2008).	Long	Hydrothermal coordination		X				multi-stage stochastic linear programming problem	SDDP
WILMAR	Meibom et al. (2011)	Short	Unit commitment and dispatch	X		X		X	stochastic mixed integer linear optimization problem	Rolling horizon
E2M2	Sun et al. (2008) / Swider and Weber (2007)	Long	Unit commitment and generation expansion planning			X		X	stochastic mixed integer linear optimization problem	Representative days, recombining three
stELMOD	Abrell and Kunz (2015)	Short	Unit commitment and dispatch			X			stochastic mixed integer linear optimization problem	Rolling horizon
VALORAGUA	Baptista and Tavares (1987)	Long	Hydrothermal coordination	X	X			X	multi-stage stochastic linear programming problem	SDP
Based on URBS	Heitmann and Hamacher (2009)	Medium	Plan generation mix			X	X		two-stage stochastic linear programming problem	Representative hours/days
SiSTEM	Mathieu et al. (2017)	Short	Unit commitment and optimal bidding in sequential markets	X		X		X	Optimization and agent-based simulation	Simulation (combining optimization and simulation)
Antares	Doquet et al. (2008), Alimou et al. (2020)	Long	System adequacy, unit commitment and dispatch	X	X	X		X	(Mixed integer) linear optimization problem	Monte-Carlo simulation and heuristics

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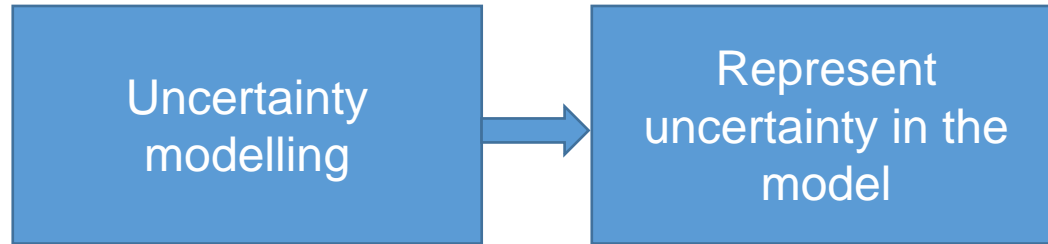
Summary

Uncertainty
modelling

- Challenges:
 - Describing the uncertainties



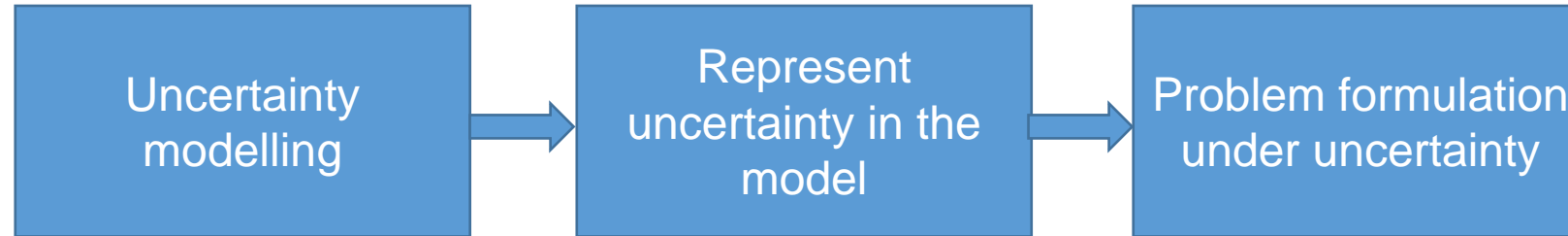
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- Challenges:
 - Describing the uncertainties
 - Making good scenarios



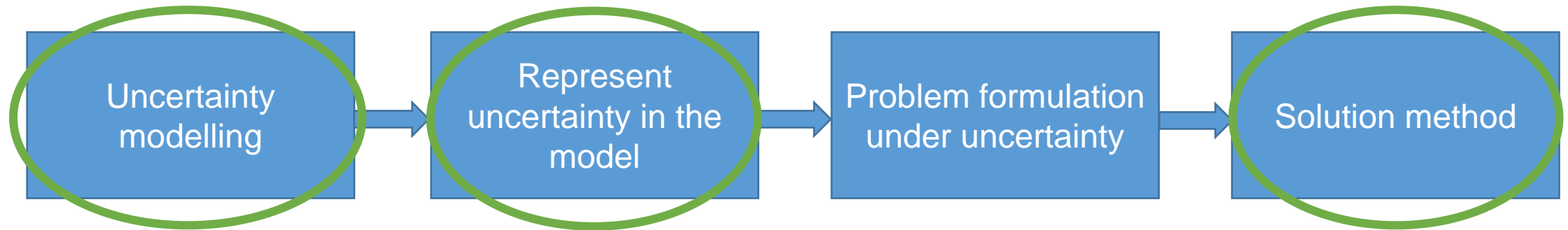
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- Challenges:
 - Describing the uncertainties
 - Making good scenarios
 - Transparency \leftrightarrow Complexity

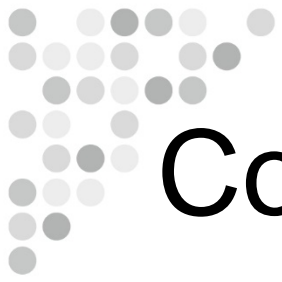


Summary



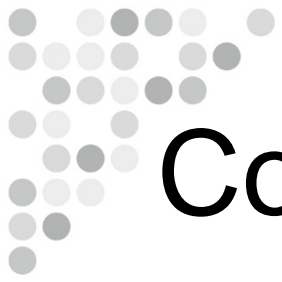
- Challenges:

- Describing the uncertainties
- Making good scenarios
- Transparency \leftrightarrow Complexity
- Computation time



Conclusions

- Stochastic optimization is the best choice!
- Wider range of uncertainties instead of more detail



Conclusions

- Stochastic optimization is the best choice!
- Wider range of uncertainties instead of more detail
- Identified research gap:

Models representing multiple dimensions of uncertainty,
and that can be solved on a real-size system in a reasonable time



Main references

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