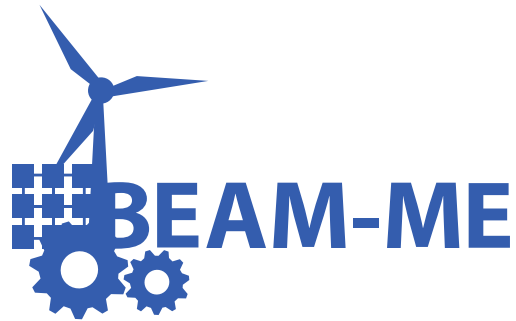




Supported by:



on the basis of a decision
by the German Bundestag

Comparison of Enhanced Benders Decomposition Methods for Stochastic Optimization of Energy System Models and Challenges for the Application to High Performance Computing

Manuel Wetzel, Frieder Borggrefe

Coimbra, 27.10.2016

DLR

A PROJECT BY

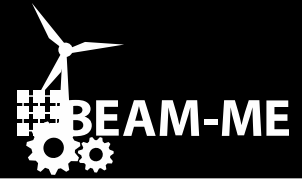


Deutsches Zentrum
für Luft- und Raumfahrt
German Aerospace Center

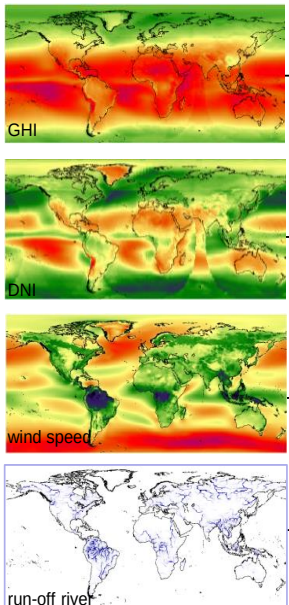
Contact:

Manuel.Wetzel@dlr.de

Energy System Model REMix at DLR



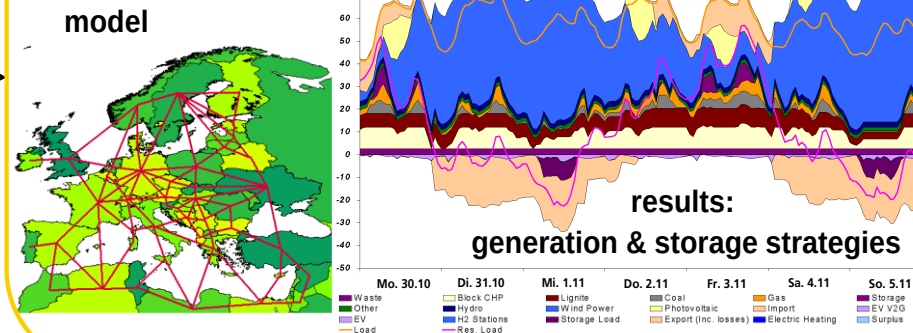
Installed capacities and power generation profiles from renewables



HVDC lines
long-range power exchange and imports

Transmission grid
based on current European AC grid

Scenario analysis with model REMix
cost minimised supply in temporal & spatial resolution



Electricity demand



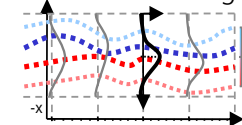
Heat demand

Flexible operation of CHP with:

- heat storages
- peak boiler & electric heaters

Electric vehicles (EV)

BEV/hybrids: charging strategies, hourly battery capacities of the fleet connected to the grid



FCEV: flexible on-site H₂ generation

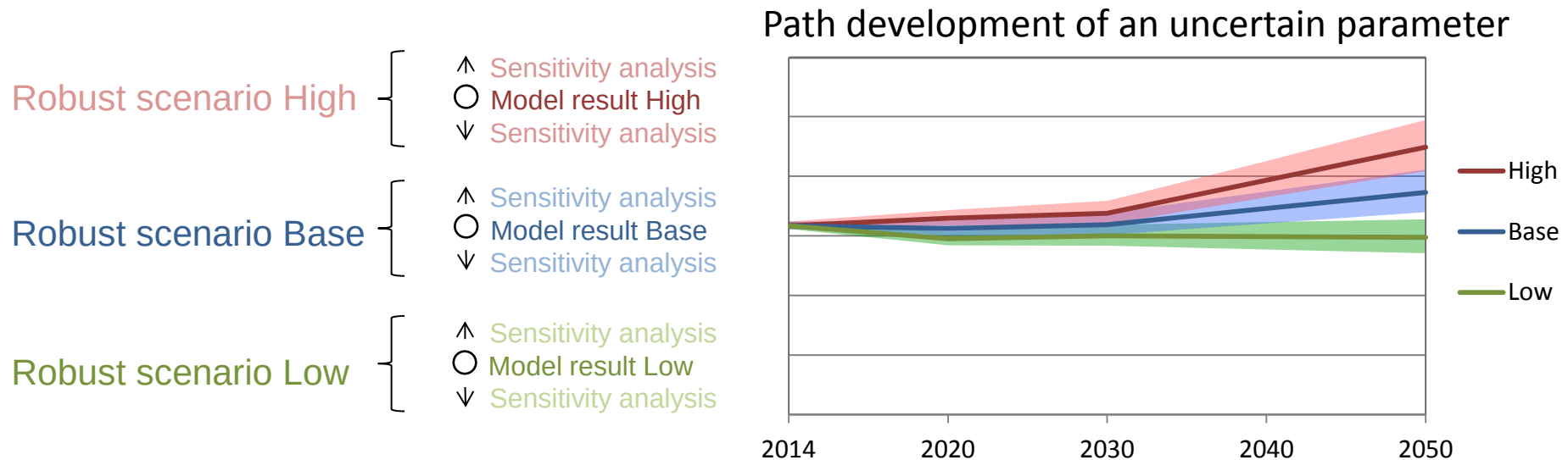
Conventional generation
nuclear, coal, gas power plants

Storages
pumped hydro
compressed air
hydrogen

Demand side management
industry & households, increases system efficiency



- German Aerospace Agency (DLR), Department for System Analysis and Technology Assessment at the Institute for Technical Thermodynamics (TT)
- Forschungszentrum Juelich (FZJ), Juelich Supercomputing Centre (JSC)
- University of Stuttgart, High Performance Computing Center Stuttgart (HLRS)
- Zuse Institute Berlin (ZIB), Department for Mathematical Optimization and Scientific Information
- Technical University Berlin, Institute for Mathematics
- GAMS Software GmbH



- Long term capacity expansion planning requires **making decisions now**, which have an impact on the energy system for several decades.
- **Context scenarios** can give a general direction of development, however a large number of **consistent sub-scenarios** are possible. The current approach usually considers only the main scenario and evaluates robustness by sensitivity analysis of sub scenarios.

Benders decomposition for two-stage stochastic optimization

(L-shaped Method)

[Benders 1962 and Van Slyke 1969]

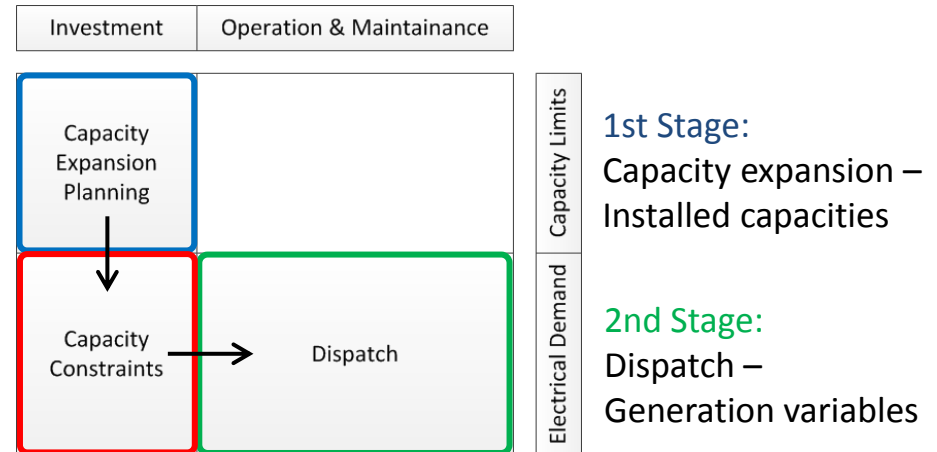
Master problem:

- **Decide now** which capacities should be expanded

Sub problem:

- **Decide later** on economic dispatch to satisfy electrical demand with capacities given by master problem

Mathematical formulation in LP table



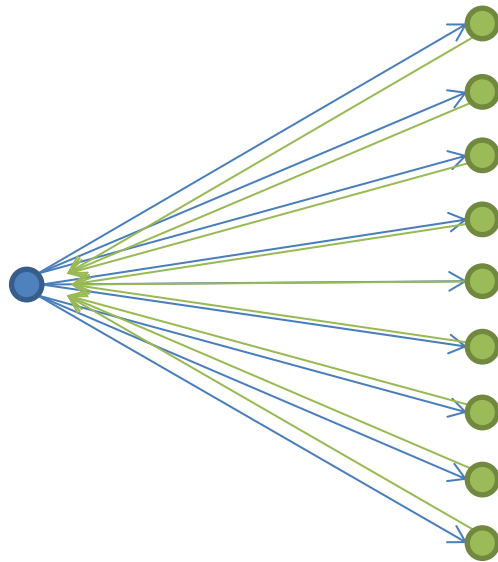
Linking variables:

Installed capacities, connecting the capacity expansion problem and the dispatch problem

1. Two-stage decision tree

Stage 1
(Master problem)
Investment decision

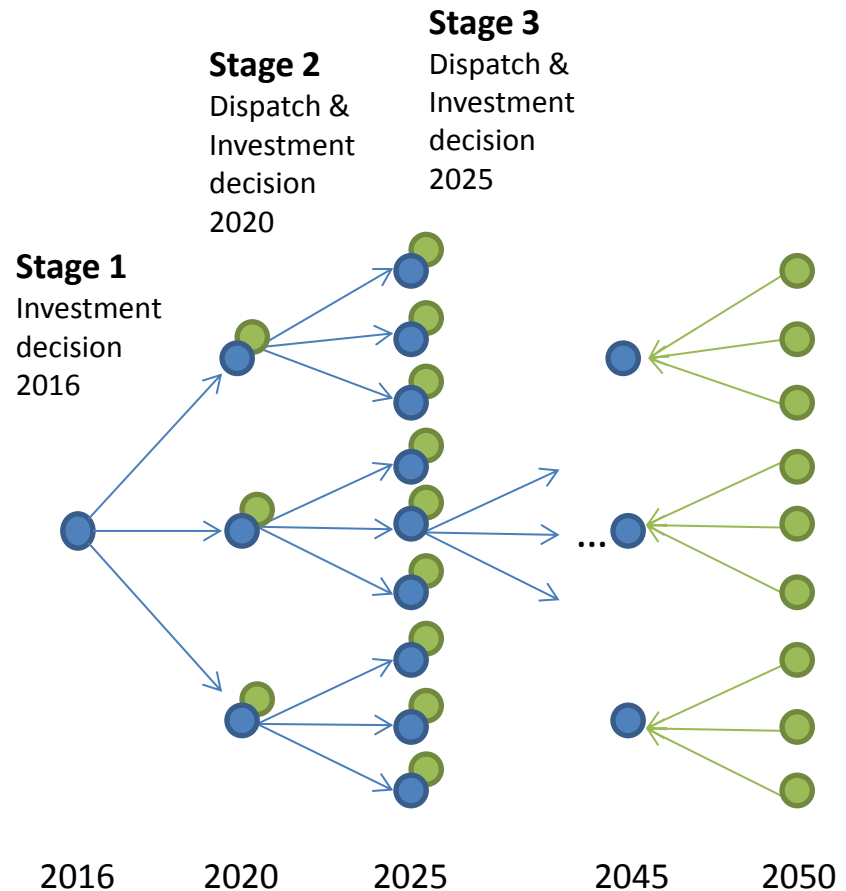
Stage 2
(Sub problem)
Stochastic economic dispatch scenarios



● Investment

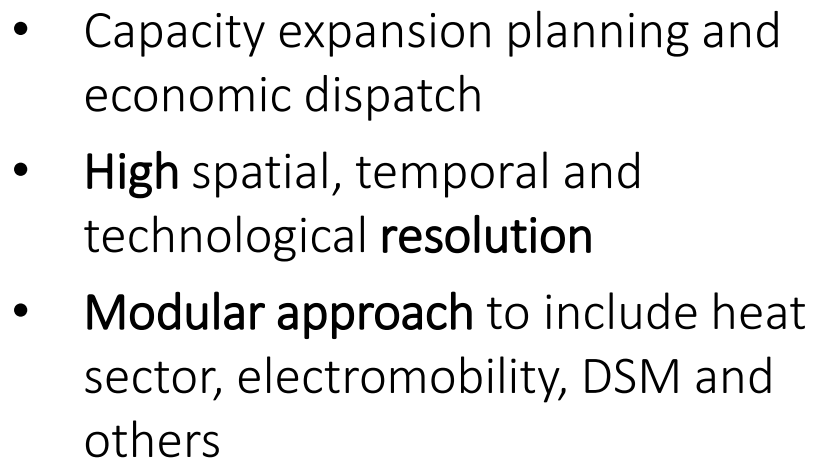
● Dispatch

2. Multi-stage decision tree



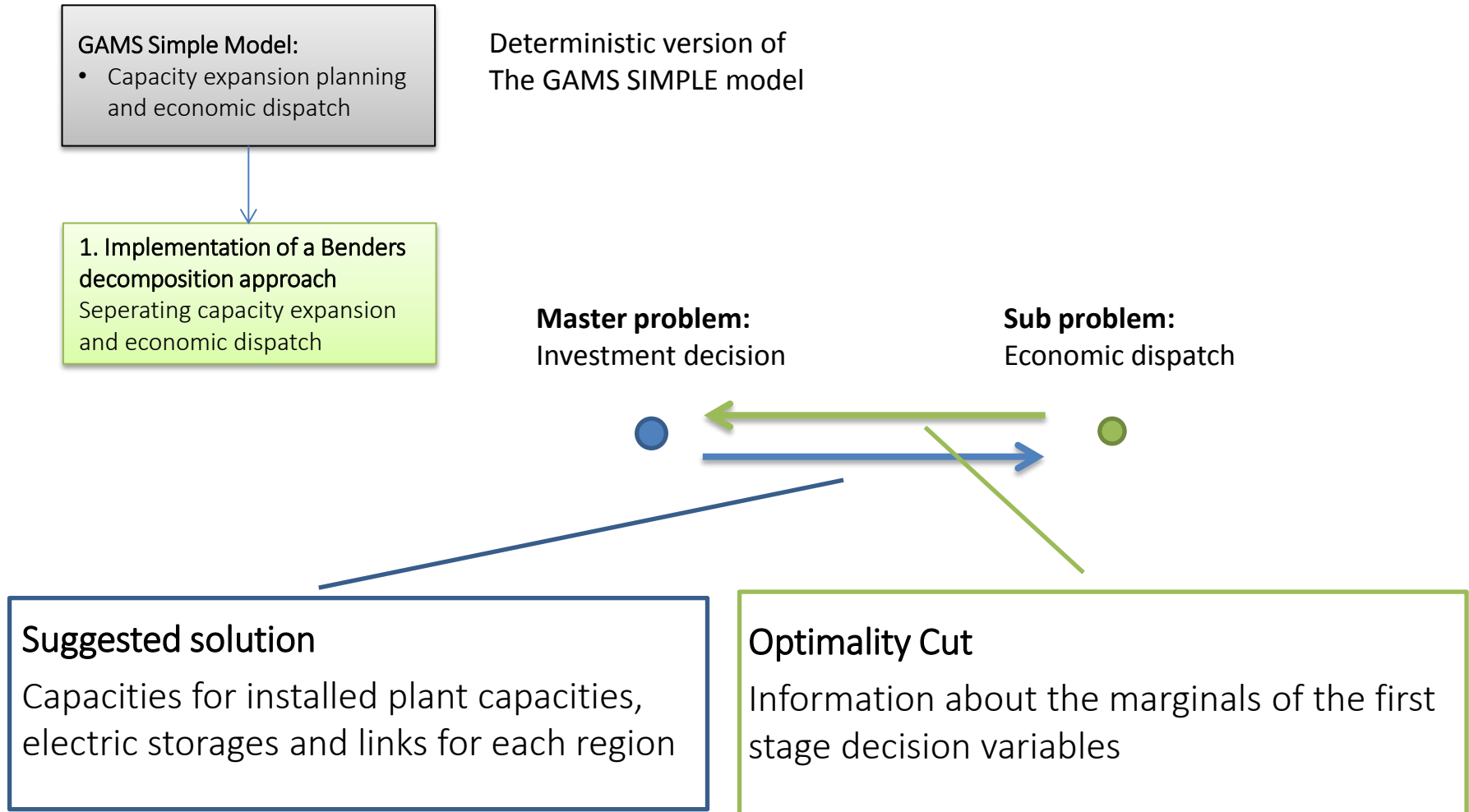


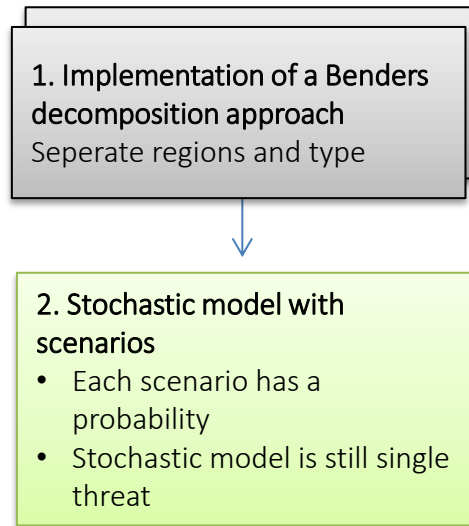
GAMS

[illegible]

- **Simplified version** of REMix, used as a development platform
- **Scalability** of model dimensions by generic data generation
- Modified for demonstrating implementation approaches for stochastic optimization

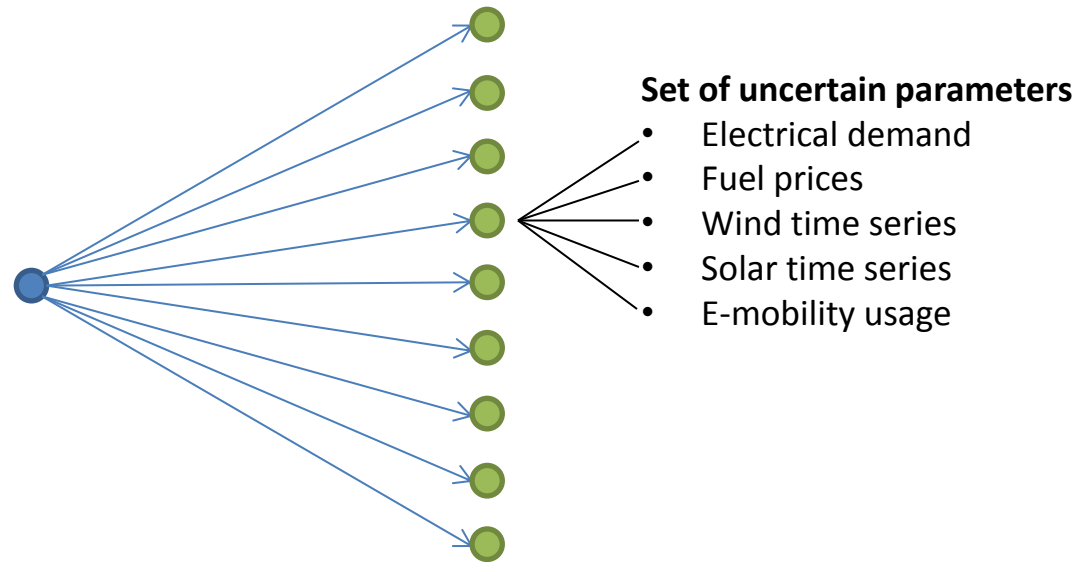
Implementation of Enhanced Benders decomposition approaches



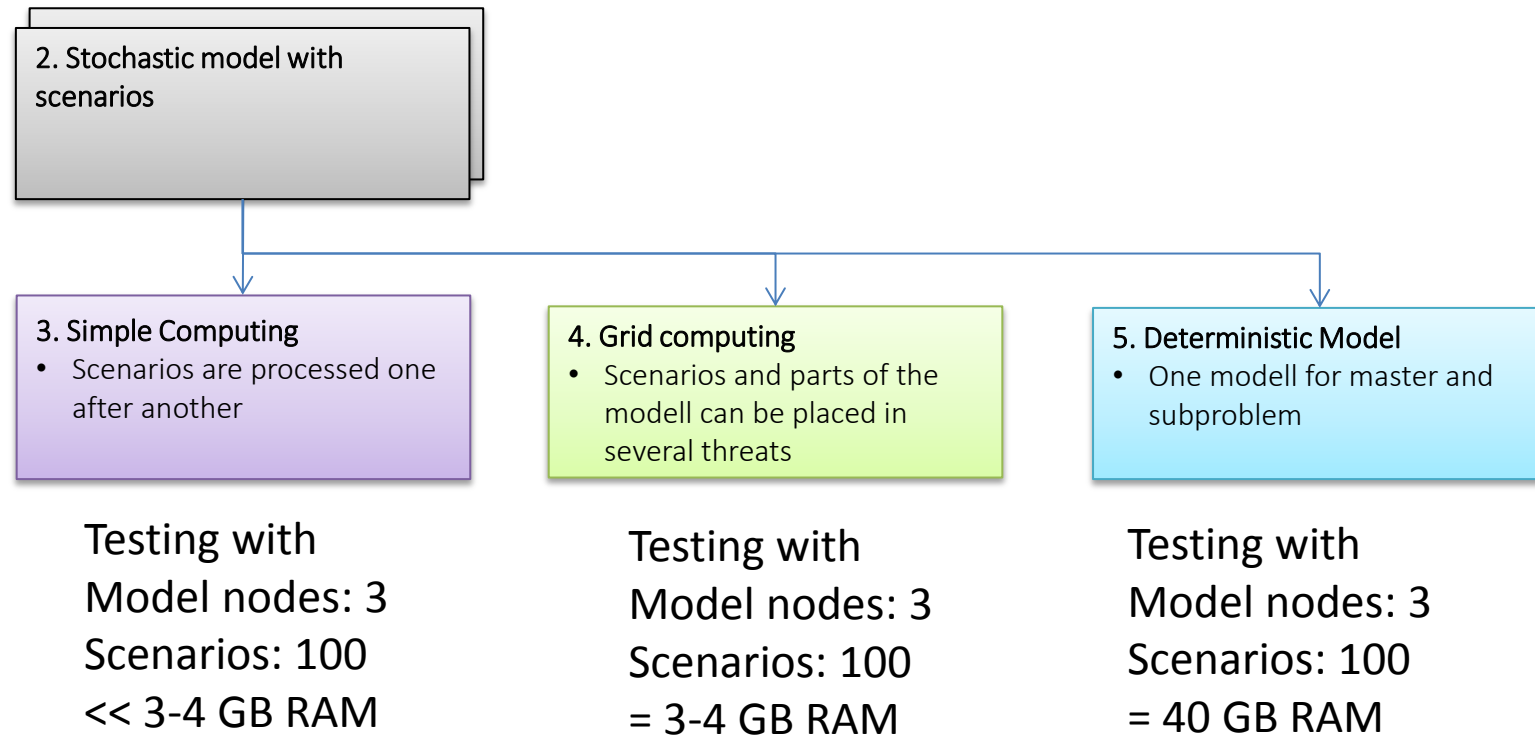


Master problem:
Investment decision

Sub problem:
Economic dispatch scenarios

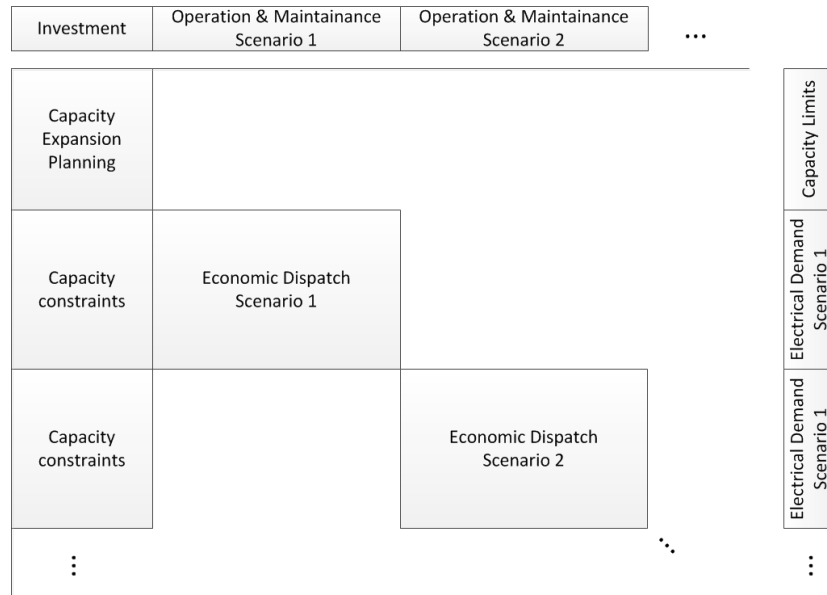


- Each subproblem is weighted by a probability and represents a specific set of assumptions for uncertain parameters



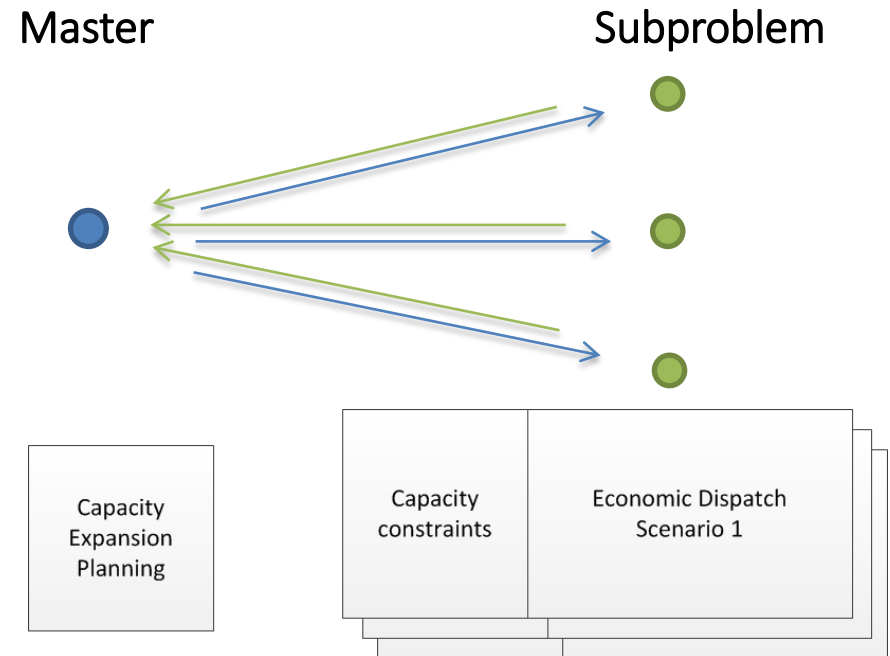
- **Small ESM** with 3 power plants, 3 storages, 2 node links and 8760h
- Formulation as Deterministic Equivalent for small problems faster than Benders decomposition, but leads to **memory restrictions** in large models

Deterministic Equivalent



- Size of LP increases with number of scenarios, solved by SIMPLEX / Barrier
- **Out of memory** for typical REMix problems with scenario dimension

Benders Decomposition



- Subproblems can be **solved in parallel**
- Memory demand **scales** with number of parallel solve processes

Improvement by optimality cut type

Singlecut

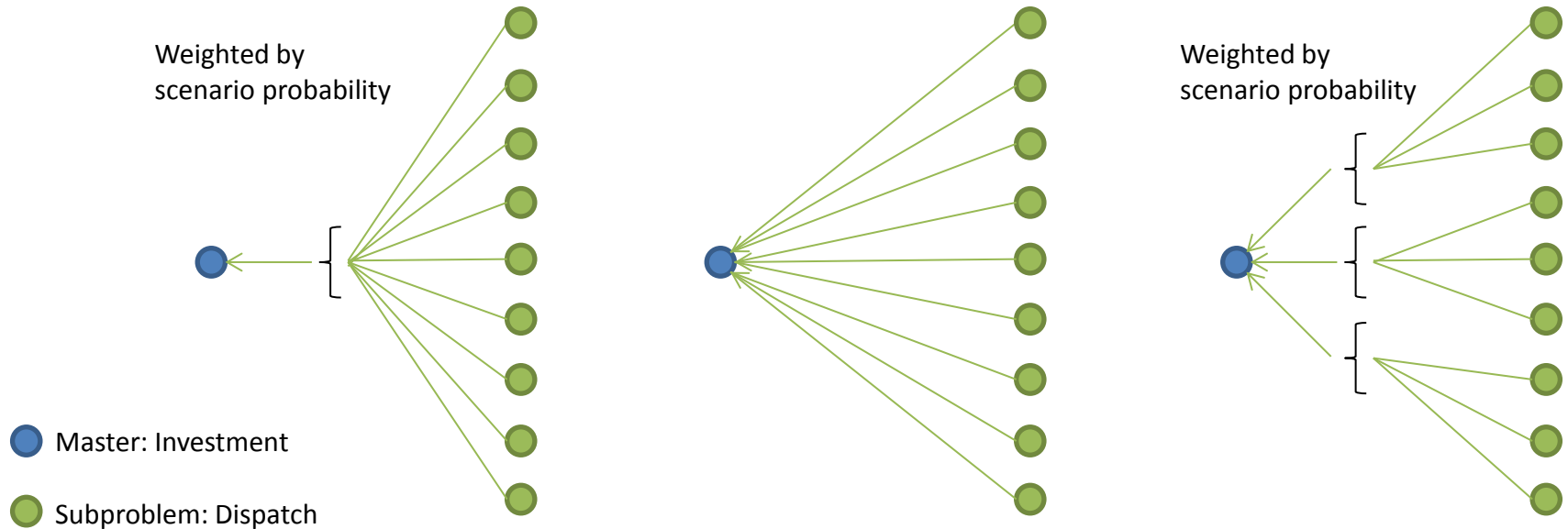
- One optimality cut per iteration is submitted to master problem

Multicut

- Each subscenario submits an optimality cut each iteration

Clustercut

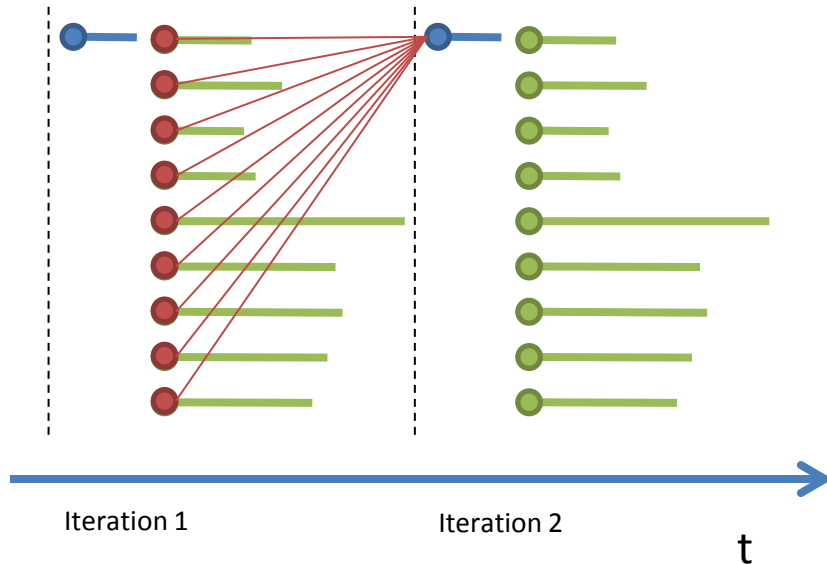
- Each cluster submits an optimality cut each iteration



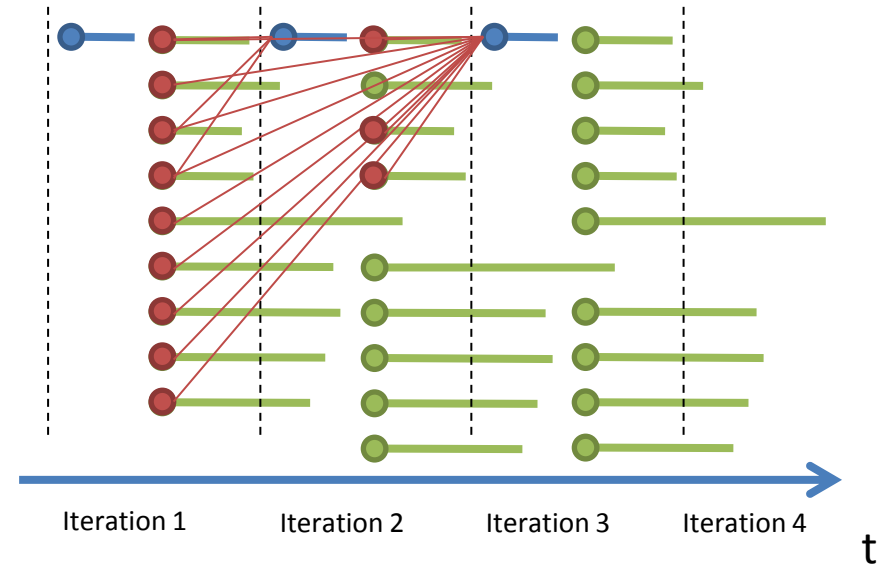
- Multicuts give **detailed information** but increase complexity [Birge 1988]
- Singlecuts **aggregate information** therefore more iterations are necessary
- Dynamic switching between cut-types is possible [Skar 2014]

Improvement to achieve asynchronity

Synchronous Model

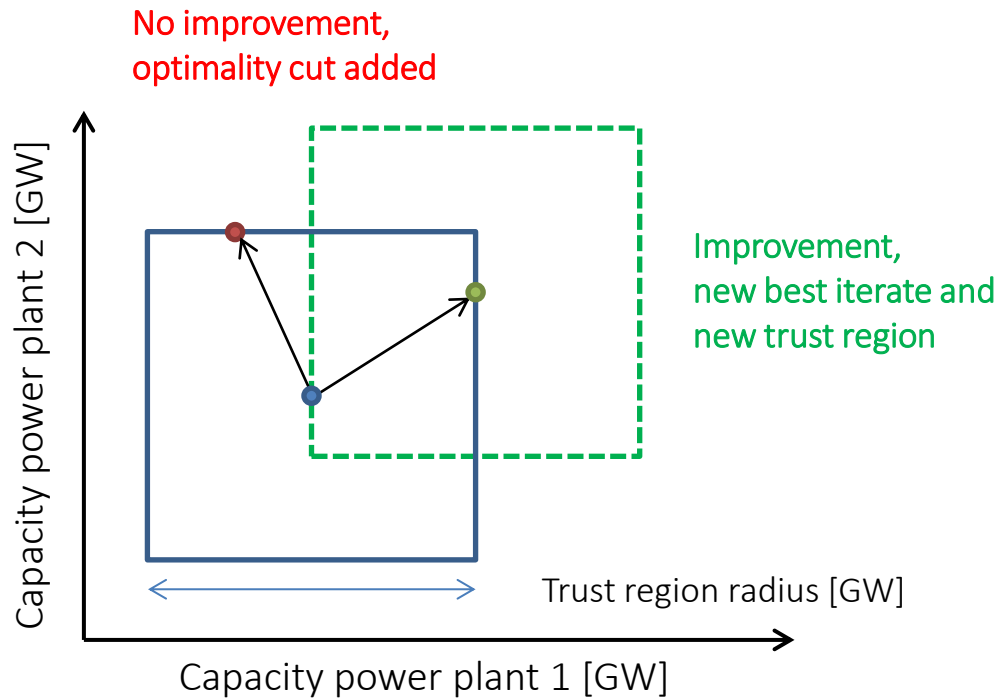


Asynchronous Model



- New Master can be started as soon as a **fraction of scenarios are solved**, new iteration has only **partial information** therefore the number of iterations increases [Linderoth 2003]
- Trade-off between more iterations and parallelisation

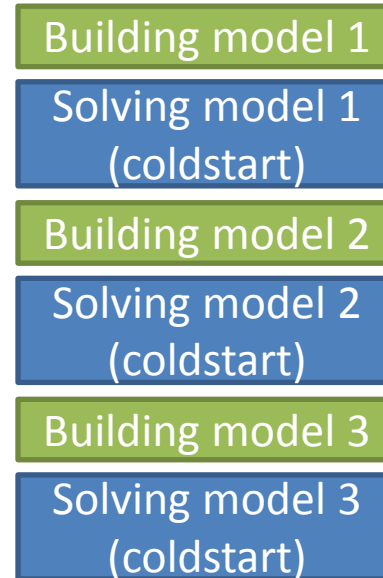
Improvement by Trust Regions



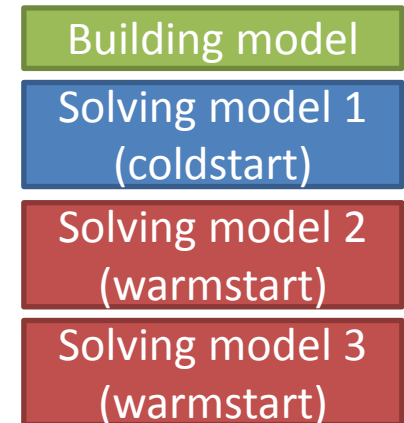
- Typical **oscillation** in Benders Decomposition between best and worst solution **can be kept low**, therefore **starting points are possible**. These can be estimated for stochastic optimization by the mean of scenario values.
- Trust regions can be **dynamically adapted**, based on performance of the new iterations.

- **Gathers** data from different sources/symbols in collection of models,
- **Updates** a base model instance with scenario data,
- **Solves** the updated model instance,
- **Scatters** the scenario results to output parameters

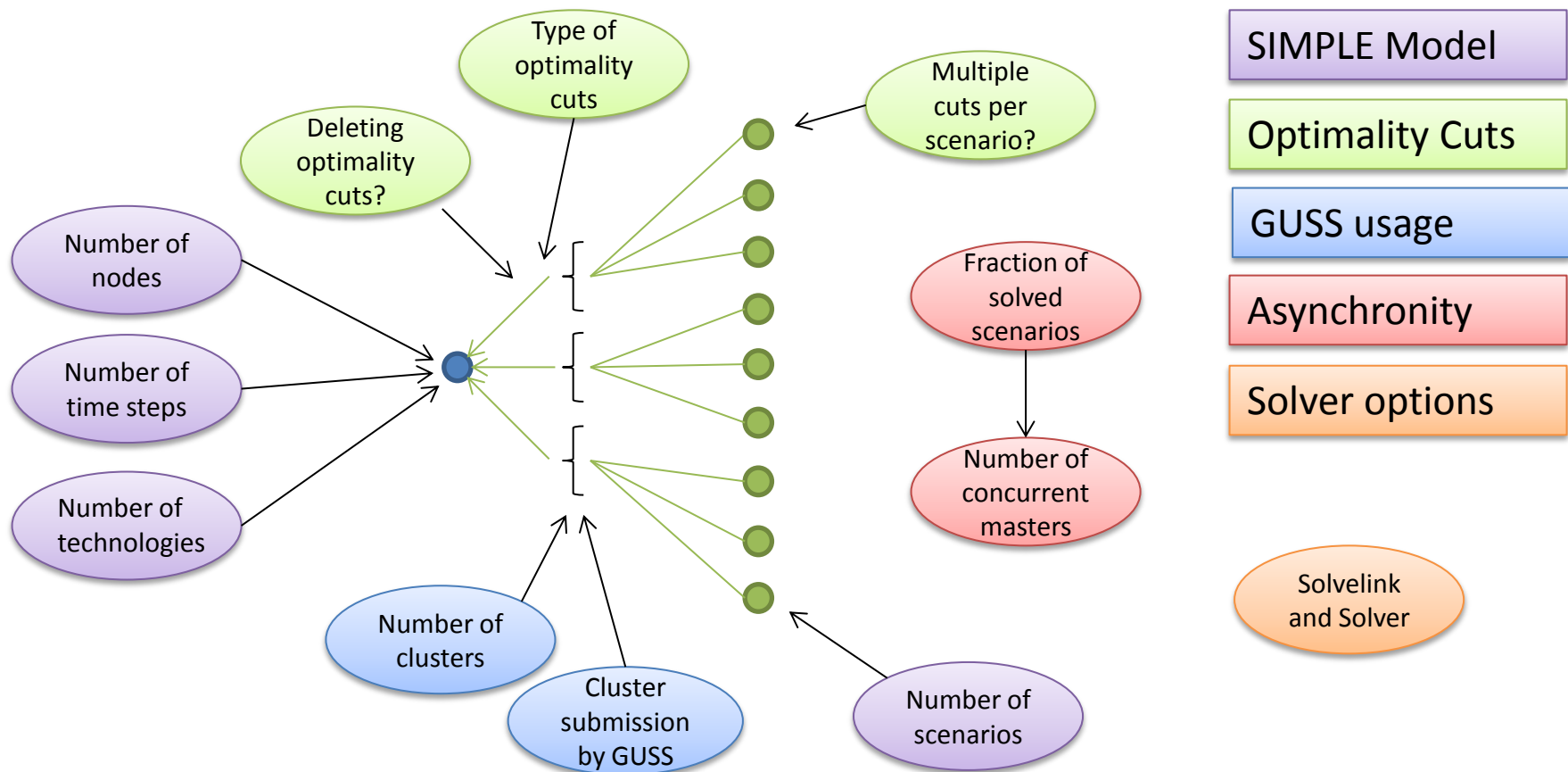
Regular



GUSS



- Model has to **built only once** and **warmstarting** improves solving time drastically, however a the models in a GUSS instance can **not be solved in parallel**, but GUSS instances can be run in parallel



- Large number of parameters to optimize performance of Enhanced Benders Algorithm making **systematic testing necessary**

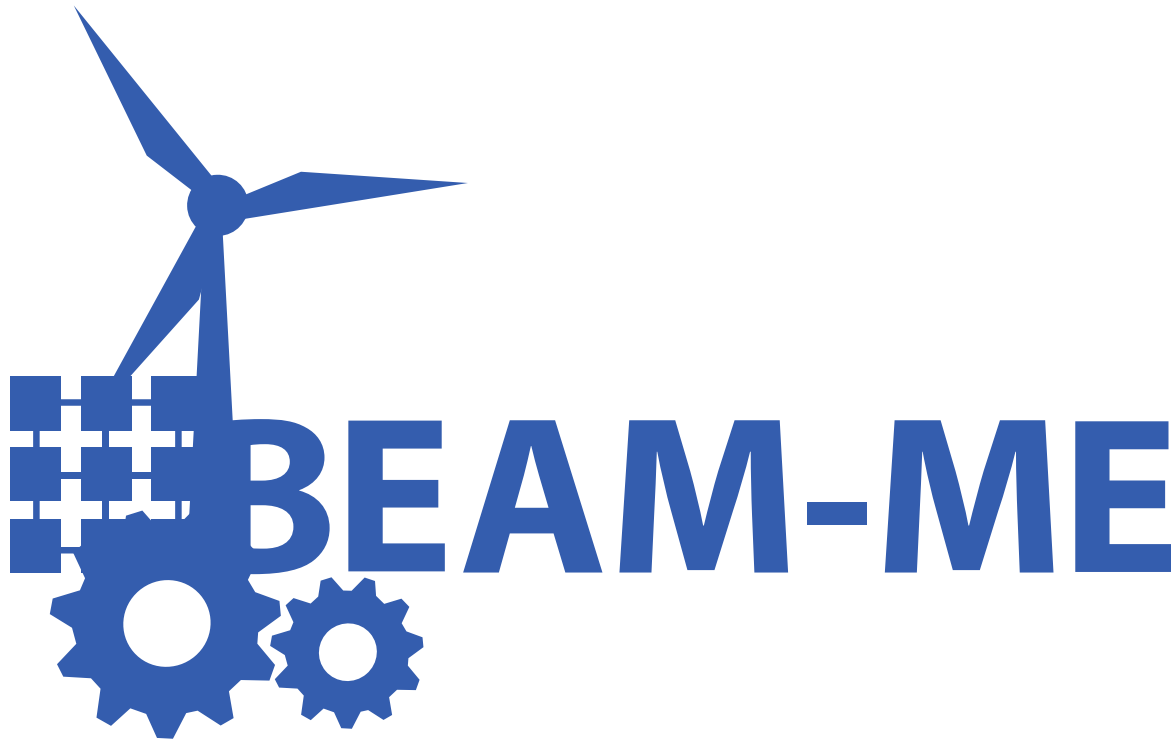
Computational results

Nodes	Scenarios	TrustRegion	Cuttype	Synchronity	GUSS usage	Wallclocktime [hh:mm:ss]	Iterations	Avg Time / Iteration [s]
3	5		Multi	Async		00:37:45	129	17.5
3	5		Multi	Sync		00:31:50	135	14.1
3	5		Multi	Sync	GUSS	00:25:40	141	10.9
3	5		Single	Sync		01:07:23	300	13.5
3	5		Single	Sync	GUSS	00:53:07	300	10.6
3	5	TR	Multi	Async		00:17:41	60	17.5
3	5	TR	Multi	Sync		00:12:33	58	12.9
3	5	TR	Multi	Sync	GUSS	00:10:44	59	10.9
3	5	TR	Single	Sync		01:02:35	300	12.5
3	5	TR	Single	Sync	GUSS	00:51:51	300	10.4
3	25		Multi	Async		07:10:55	143	187.8
3	25		Multi	Sync		01:52:12	118	57.0
3	25		Multi	Sync	GUSS	01:04:10	130	29.6
3	25		Single	Sync		05:06:43	300	61.3
3	25		Single	Sync	GUSS	02:33:43	300	30.7
3	25	TR	Multi	Async		01:19:11	55	85.6
3	25	TR	Multi	Sync		00:49:29	53	56.0
3	25	TR	Multi	Sync	GUSS	00:26:14	53	29.6
3	25	TR	Single	Sync		04:30:07	300	54.0
3	25	TR	Single	Sync	GUSS	02:30:21	300	30.1

- All configurations with 8 model nodes instead of 3 were **aborted** due to the 300 iterations limit

- **Reducing computational overhead** by using methods which simplify or reduce model building time (Usage of GUSS, externalizing model building)
- **Balancing CPU load** for migration to high performance computing
- **Generating better cuts** - insufficient electrical generation will cause all plant types in a region to be built
- **Deleting unused cuts** in order to keep the masterproblem as small as possible while retaining important information
- **Improving the starting point** for the Trust-Region approach
- Combining stochastic optimization with **decomposition on the scenario level** (decomposition in model nodes or timesteps) in a Nested Benders approach

Project BEAM-ME



Supported by:



Federal Ministry
for Economic Affairs
and Energy

on the basis of a decision
by the German Bundestag

A PROJECT BY



Deutsches Zentrum
für Luft- und Raumfahrt
German Aerospace Center



- J. Benders, "Partitioning procedures for solving mixed variables programming problems," *Numerische Mathematik*, vol. 4, pp. 238–252, 1962.
- R. M. Van Slyke and R. Wets, "L-shaped linear programs with applications to optimal control and stochastic programming," *SIAM Journal on Applied Mathematics*, vol. 17, no. 4, pp. 638–663, 1969.
- J. R. Birge and F. V. Louveaux, "A multicut algorithm for two-stage stochastic linear programs," *European Journal of Operational Research*, vol. 34, no. 3, pp. 384–392, 1988.
- C. Skar, G. Doorman and A. Tomasgard, "Large-scale power system planning using enhanced Benders decomposition," *Power Systems Computation Conference (PSCC)*, pp. 1-7, 2014,, doi: 10.1109/PSCC.2014.7038297
- J. T. Linderoth and S. J. Wright, "Implementing a decomposition algorithm for stochastic programming on a computational grid," *Computational Optimization and Applications*, vol. 24, pp. 207–250, 2003.