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VIRTUAL 2022 - LATIN AMERICA & CARIBBEAN

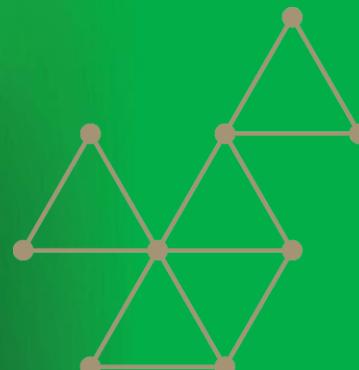


Stochastic co-optimization of renewable-driven reserves, storage and flexibility in power system planning and operation

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PSR

October 20th, 2022



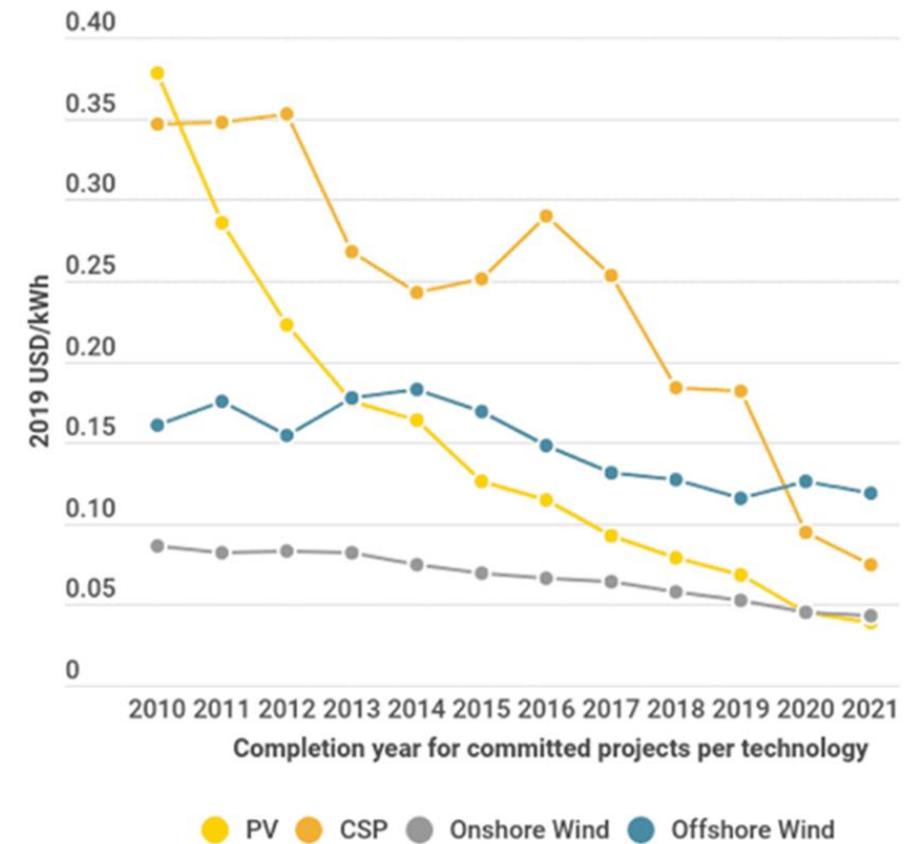
Introduction

Decarbonization: the renewable revolution

- ▶ Initial motivation: **energy policy** and climate change
- ▶ Currently: **economically motivated**, cheaper expansion



Costs continue to fall for solar and wind power technologies



Fuente: IEA & Irena (2016, 2019)

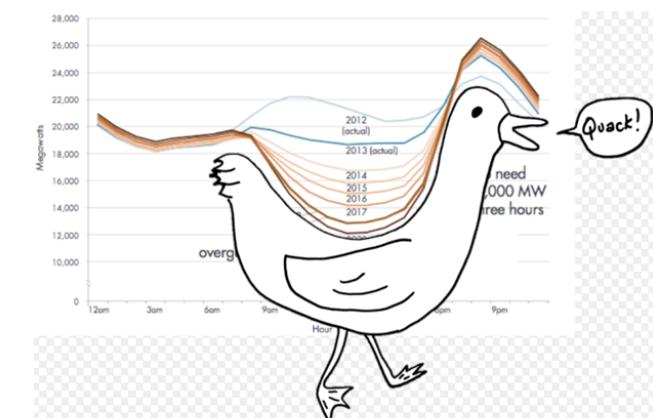
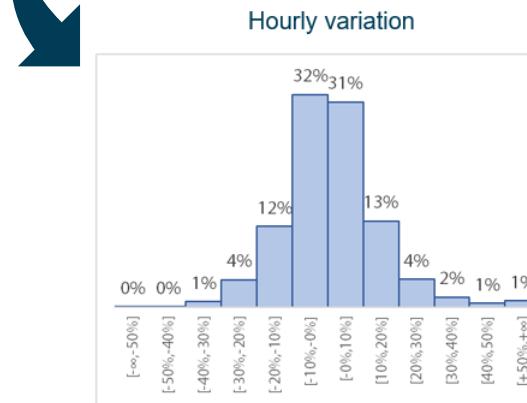
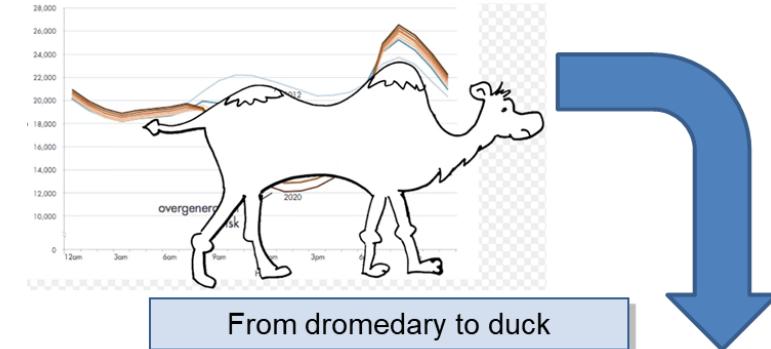
Great modelling challenges

- ▶ Need to combine long-term view of expansion with short-term view of operation

- Hourly time steps (or even smaller)
- Unit commitment decisions
- Startup costs
- Minimum down-time and up-time
- Ramping constraints
- Hydraulic constraints in river basin
- Variability of renewables, inflows and demand (uncertainties)
- Energy, capacity and reserve requirements

Statistical tools are essential to represent the uncertainty and variability of the renewables

Hourly generation of a wind plant in Brazil



- **Operating Reserve:** needed to guarantee security of supply

The “new world” of the 3Ds: decarbonization, decentralization and digitization

- Generators located in the distribution
- Consumers provide generation services
- Multiplicity of new actors
- Consequences:
 - Change in the operation of the distribution network
 - Need for regulatory adjustments

Electric Power Systems Research
Volume 211, October 2022, 108445

A methodology for improved TSO-DSO coordination in grid operation planning

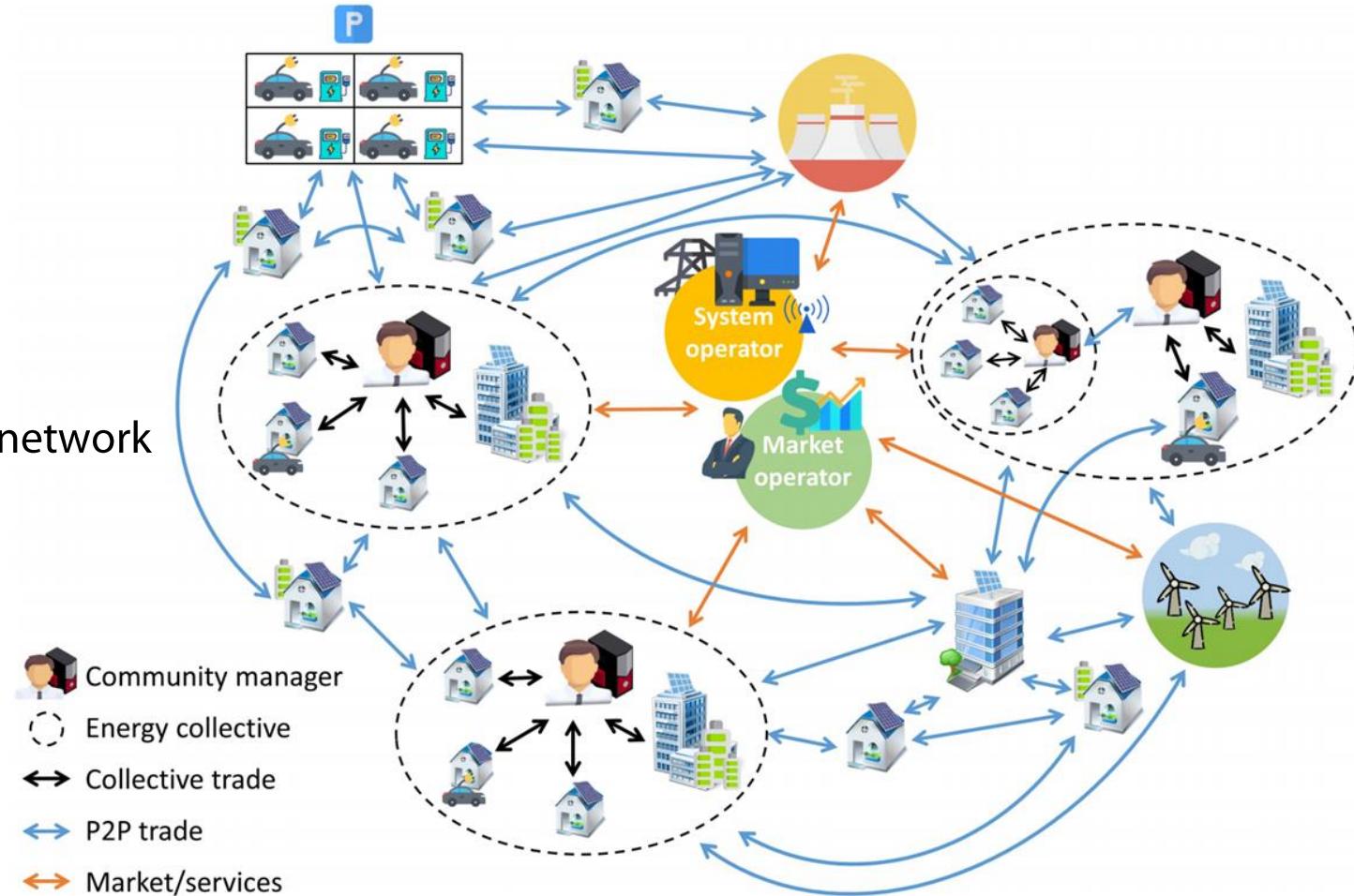
Jan Ringelstein ^a , Mike Vogt ^a, Ataollah Moghim Khavari ^b, Roberto Ciavarella ^c, Marialaura Di Somma ^c, Giorgio Graditi ^c

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Strong wholesale and retail integration



Stochastic modeling of renewables

Stochastic modeling of renewables

Data

Simulate renewable historical with reanalysis data

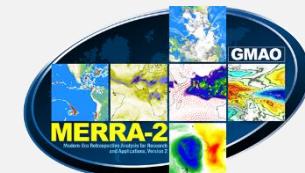
Scenarios

Stochastic modeling of renewables and hydro (with spatial correlation)

Input data



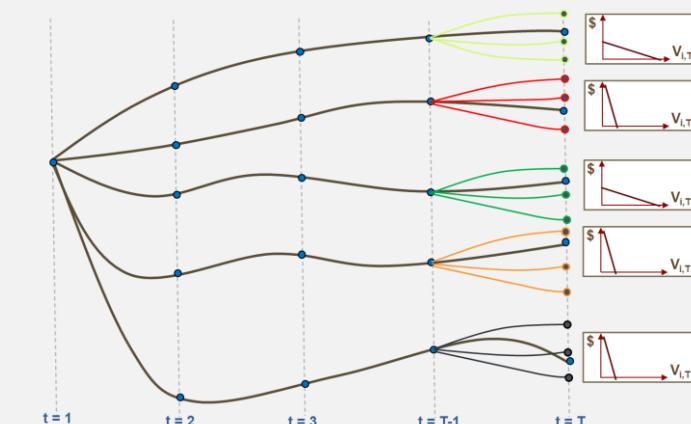
Renewable Modeling

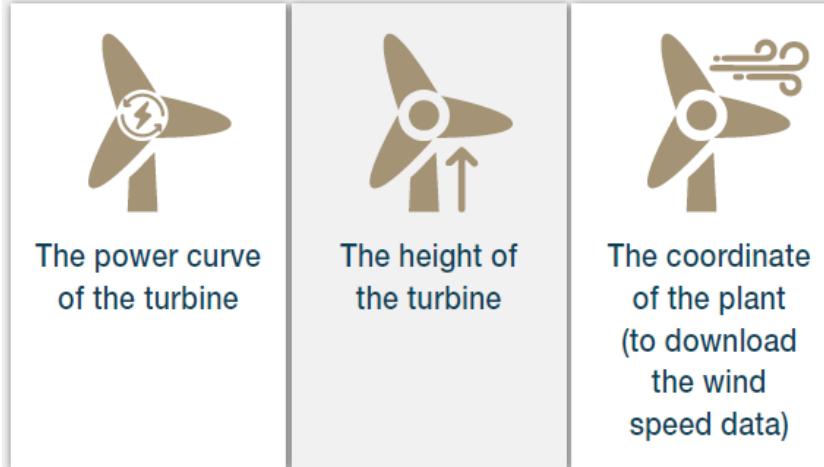


Renewable generation scenarios

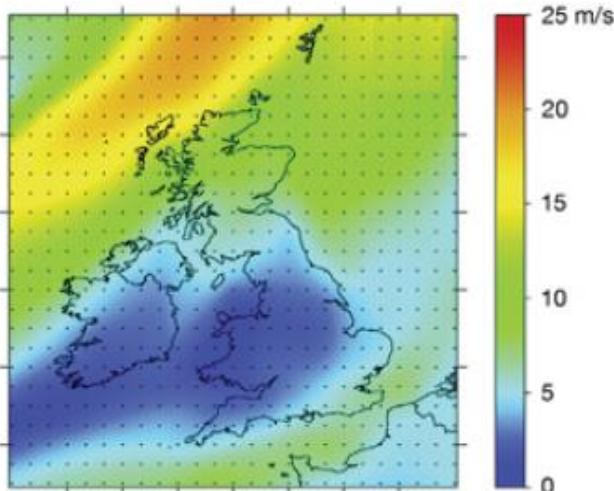


Stochastic Production Costing

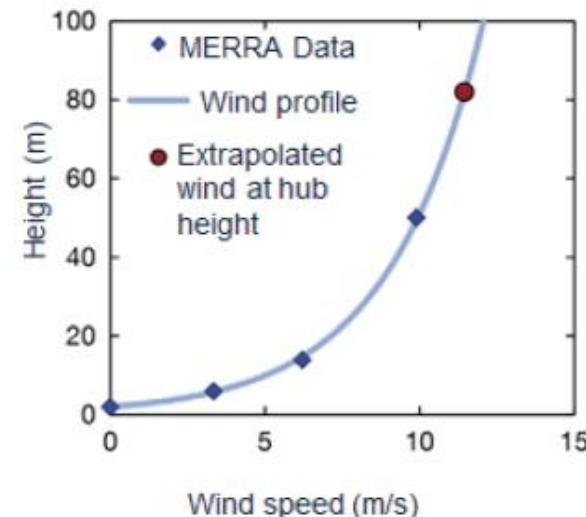




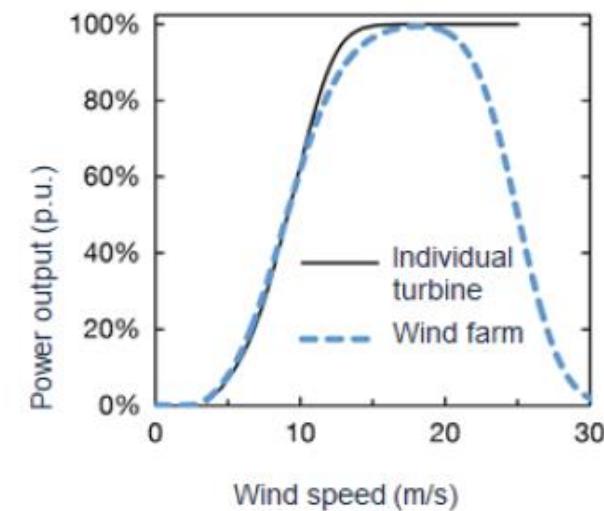
DOWNLOAD WIND SPEED DATA

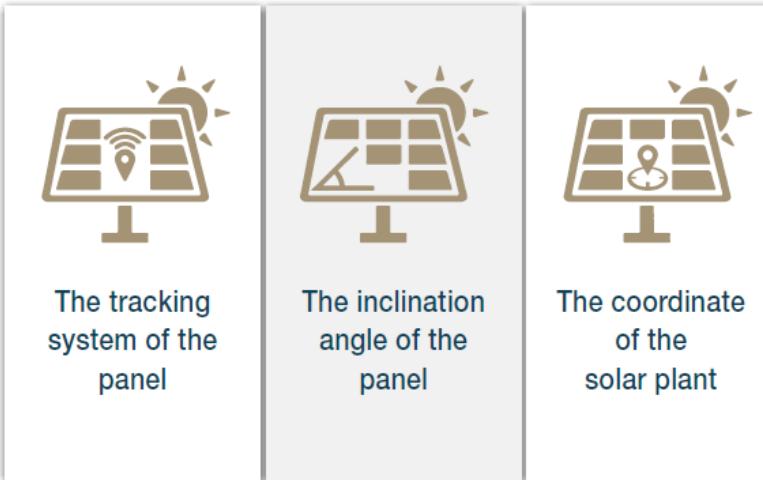


EXTRAPOLATE WIND SPEED TO THE PLANT HEIGHT

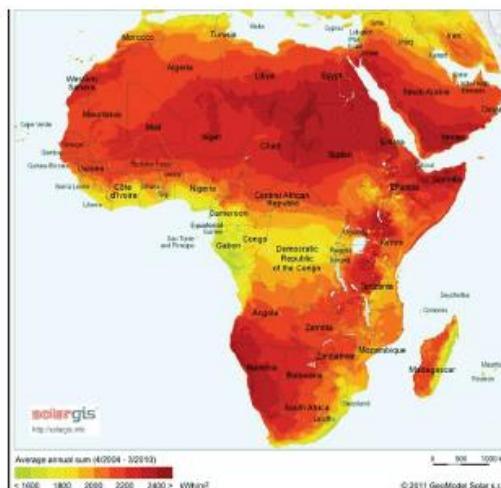


CALCULATES THE POWER OUTPUT USING THE TURBINE POWER CURVE

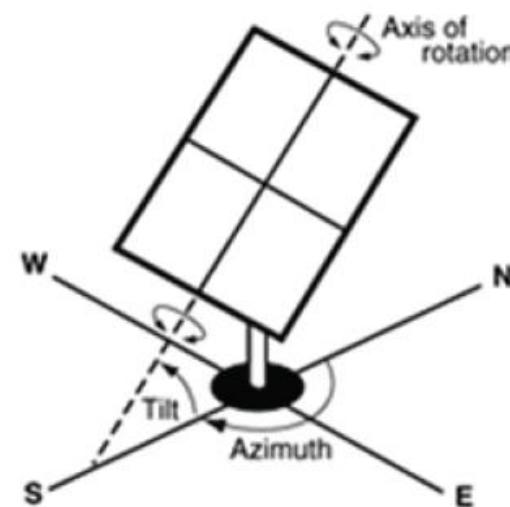




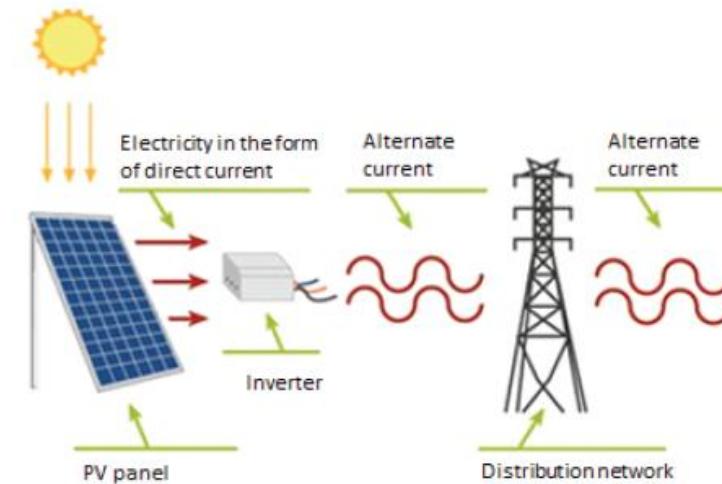
DOWNLOAD SOLAR IRRADIATION DATA



ESTIMATE POWER OUTPUT CONSIDERING EFFICIENCY AND TRACKING SYSTEM



DC → AC CONVERSION

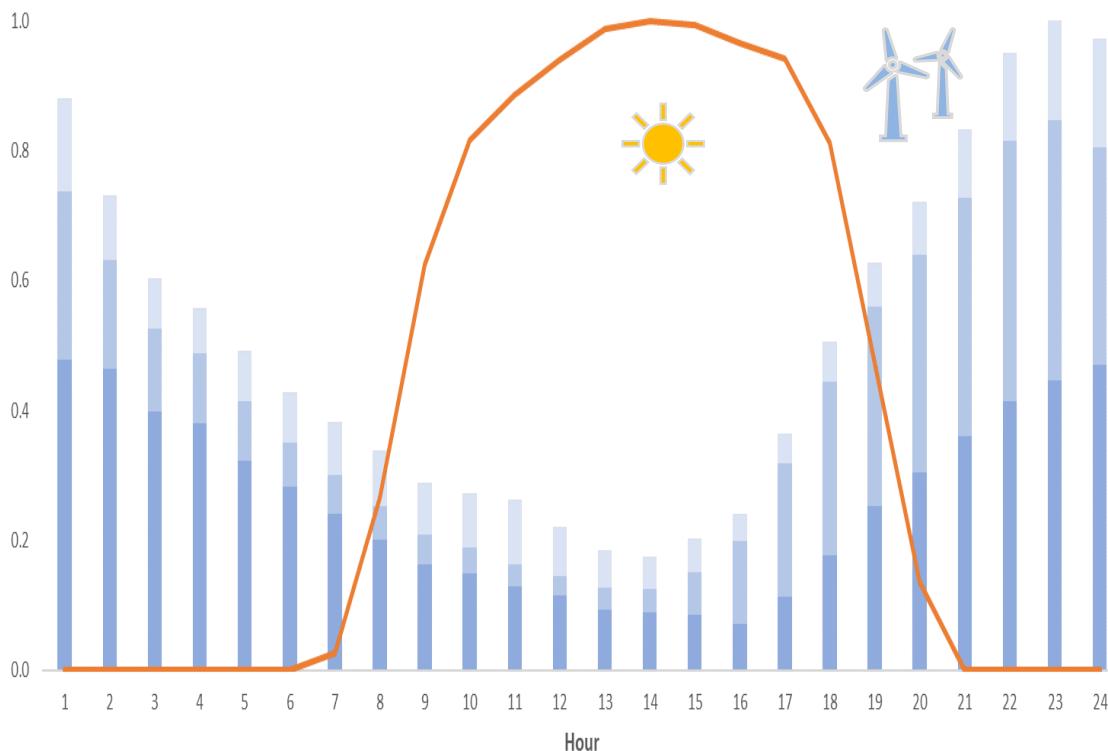


Capturing spatial & temporal correlations

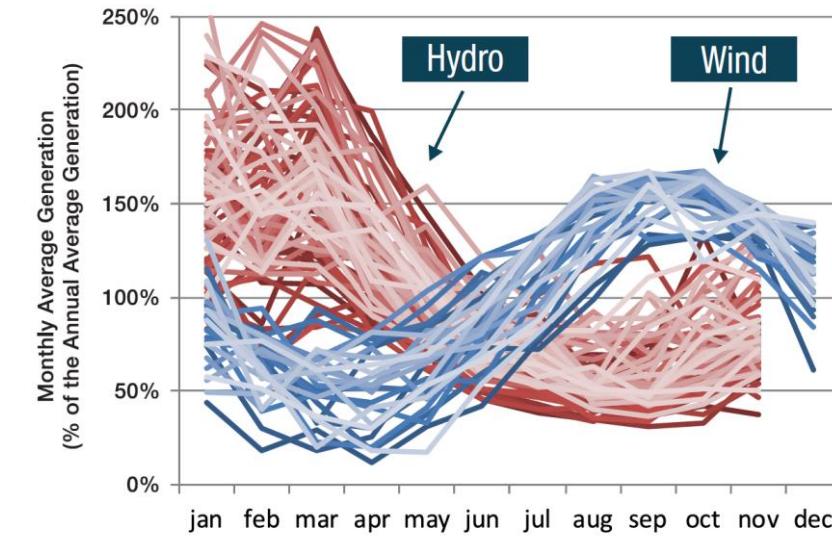
Intraday

Hourly complementarity and correlation between renewables

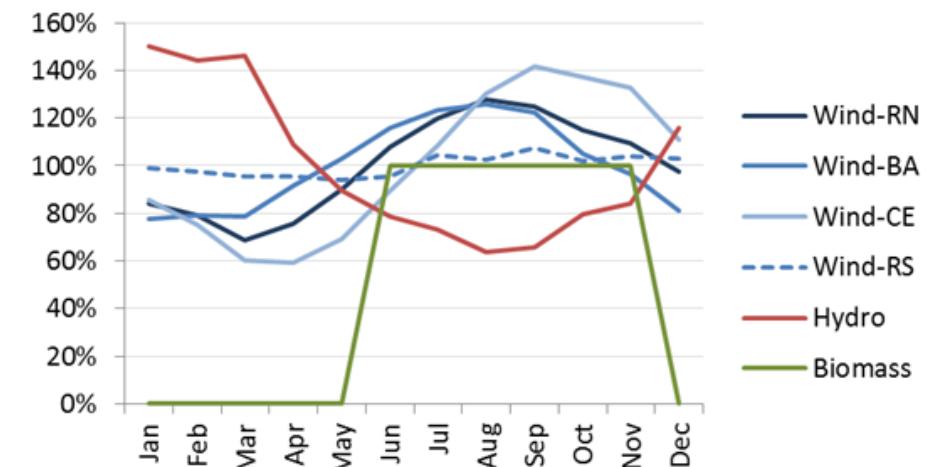
Reina Cururos Vientos Solar



Monthly complementarity and correlation between renewables



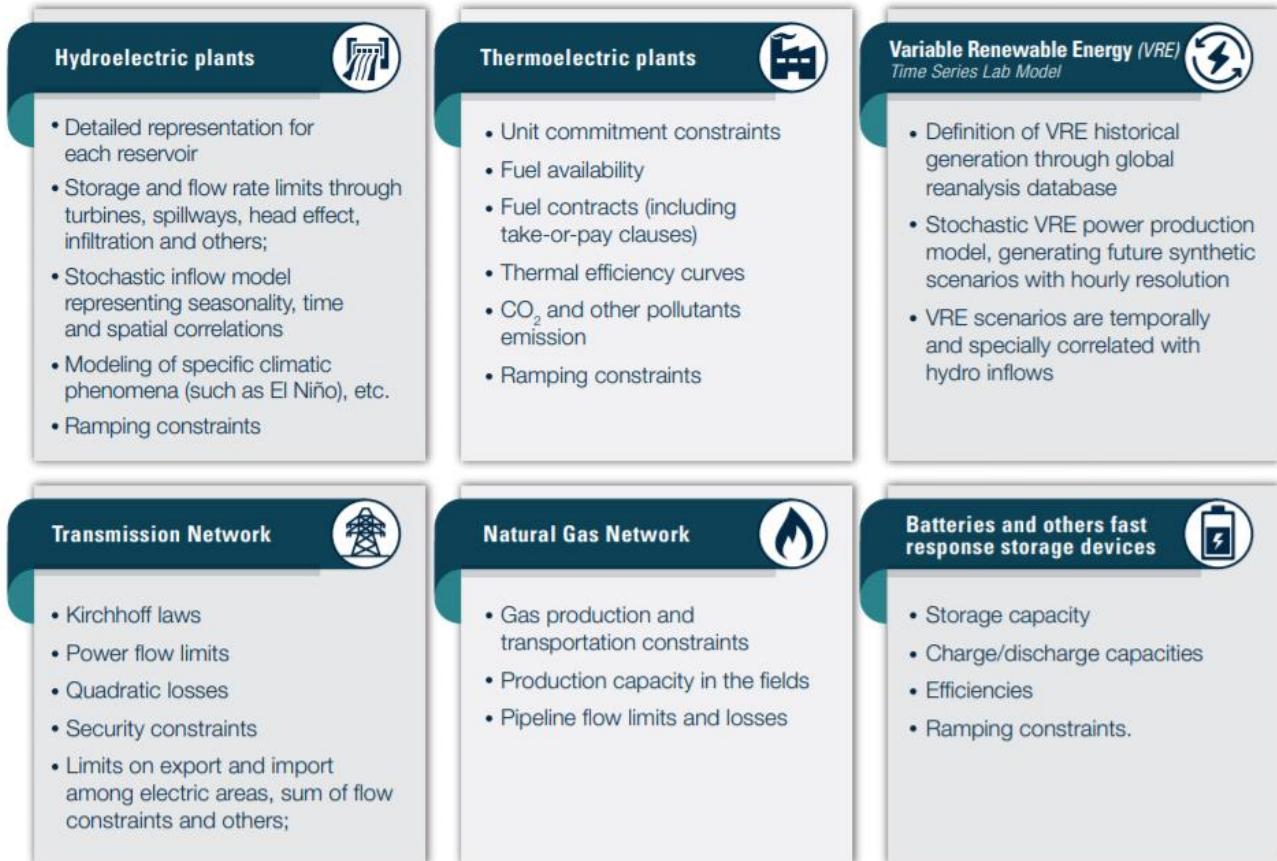
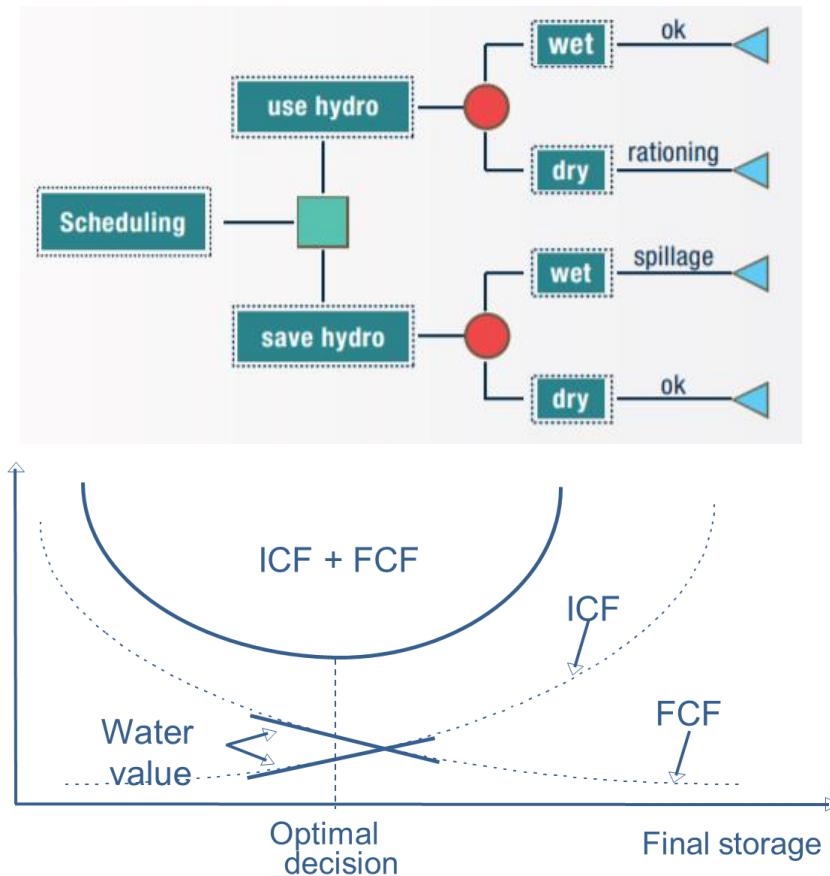
Seasonal generation patterns





Multistage stochastic optimization methodologies for electrical system planning and operation

Stochastic production costing simulations: SDDP Algorithm



- The objective of this tool is to minimize the sum of costs for purchase and transportation of fuels for thermal plants, pollutant emission costs, hydro and thermal plants' O&M costs, transmission wheeling rates, cost of energy not supplied and other penalties.

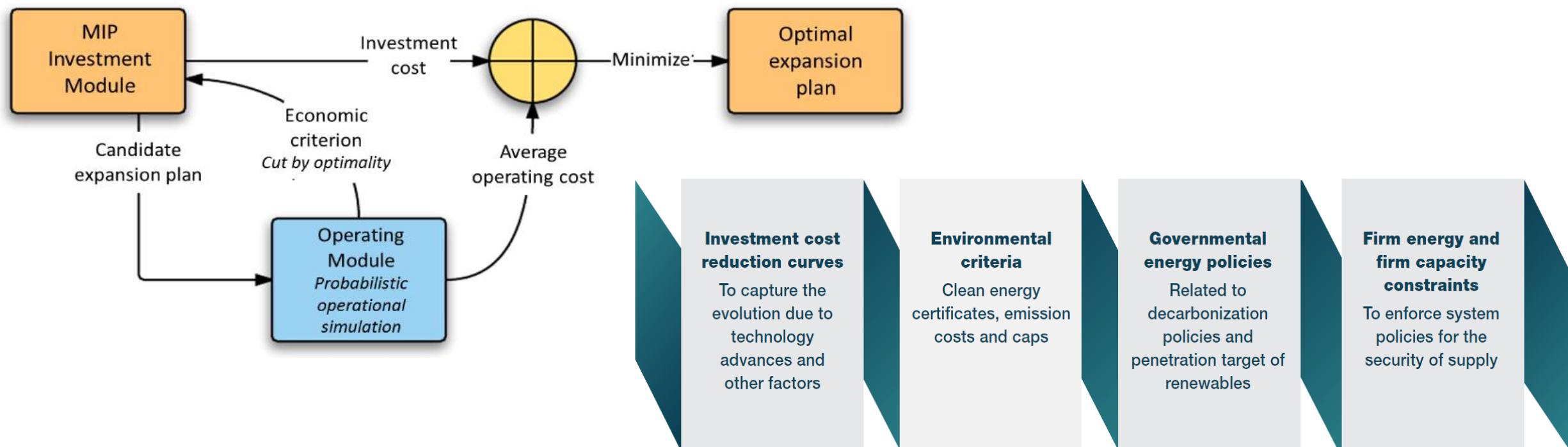
SDDP Algorithm: uncertainties that can be represented

- Hydro inflows and renewable generation (wind, solar, biomass etc.)
 - Multivariate stochastic model (PAR(p))
 - Inflows: macroclimatic events (El Niño, La Niña), snowmelt and others
 - Climate change impact
 - Spatial correlation of wind, solar and hydro
- Load uncertainty
 - Annual load growth rates
 - Markov chains/AR model
 - Demand scenarios
 - Within-stage variability
 - Monte Carlo sampling
- Uncertainty on fuel costs
 - Fuel cost scenarios
 - Markov chains
 - (hybrid SDDP/SDP model)
 - Wholesale energy market prices
 - Markov chains
 - G&T equipment outages
 - Monte Carlo sampling

Solution Algorithm:
Stochastic Dual Dynamic Programming (SDDP) →
global industry standard, with over 1.200 citations in
the scientific/engineering literature

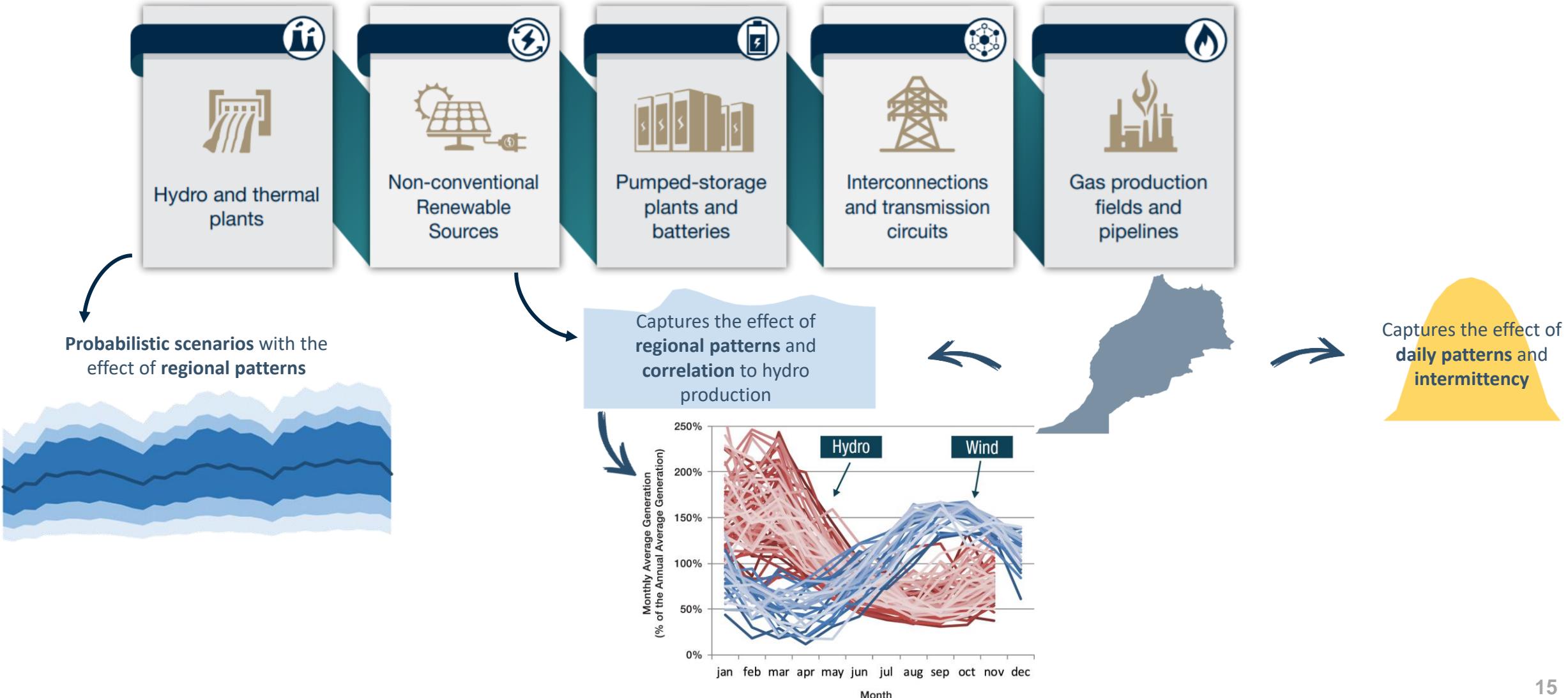
Multistage stochastic capacity expansion planning

- ▶ This tool determines the least-cost sizing and timing decisions for construction, retirement and reinforcement of generation capacities, transmission network and natural gas pipelines.
- ▶ This model optimizes the **trade-off** between **investment costs** to build new projects and the expected value of **operative costs**



Multistage stochastic capacity expansion planning

- Several types of projects are available to be contemplated by the model:

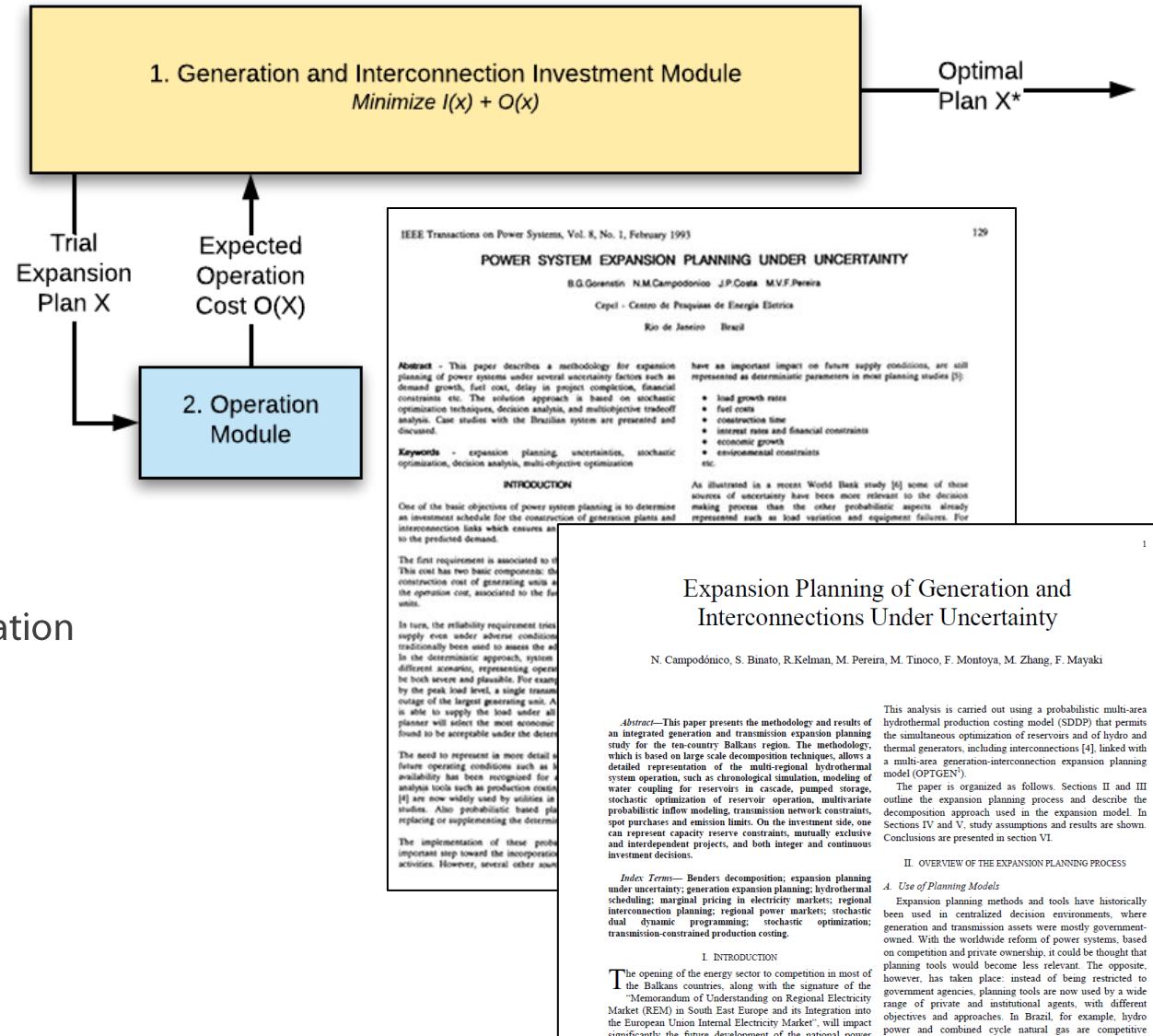




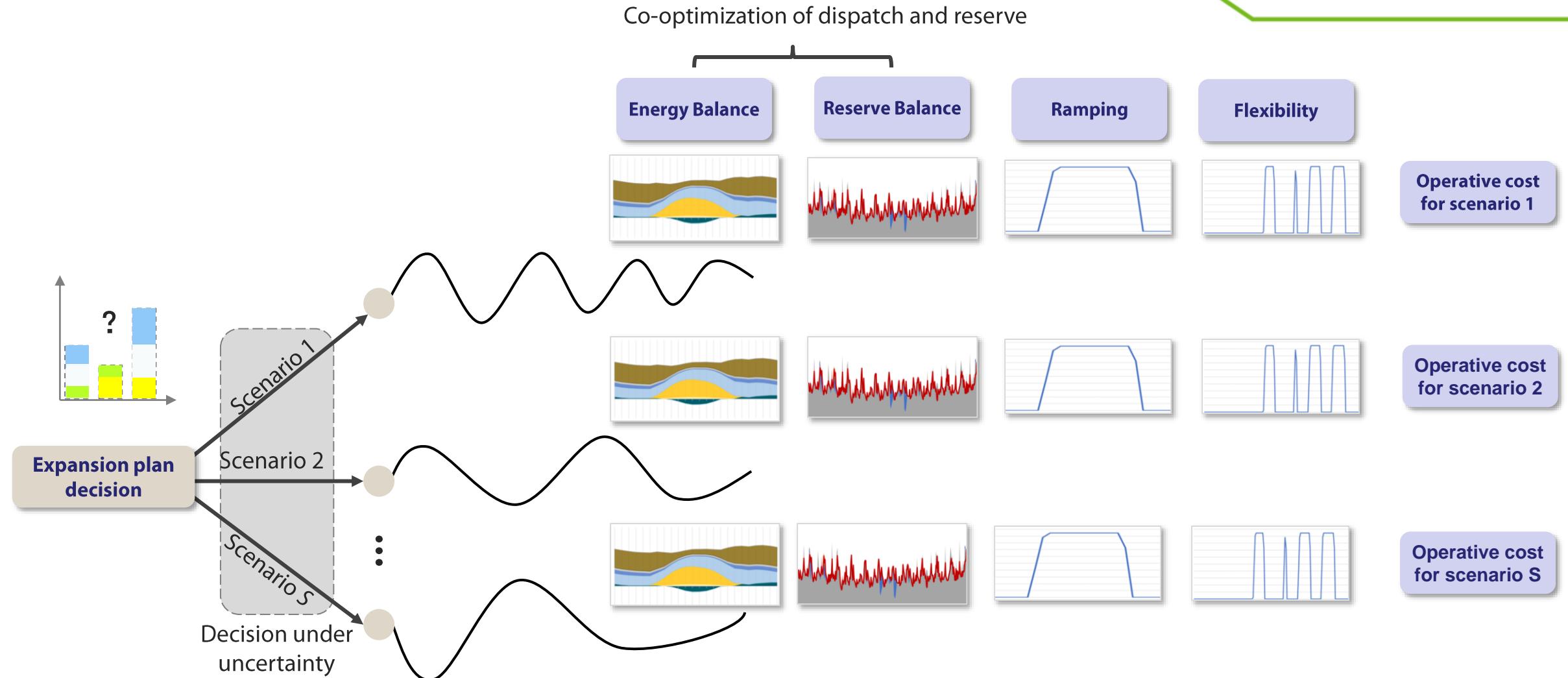
How to ensure efficiency, reliability, flexibility, and resilience in the energy transition

Planning 1.0: Economic Efficiency + Energy Policy Guidelines

- ▶ Minimize total costs to the consumer (investment + expected value of the operating costs)
 - Renewable variability (hydroelectric, wind, biomass, solar), demand (e.g., effect of temperature etc.)
- ▶ Energy policy guidelines
 - Nuclear programs, etc.
 - “New old criteria”: CO₂ emission budgets, renewable penetration targets, etc.
 - Desired energy mix



Example 1.0: Business as Usual

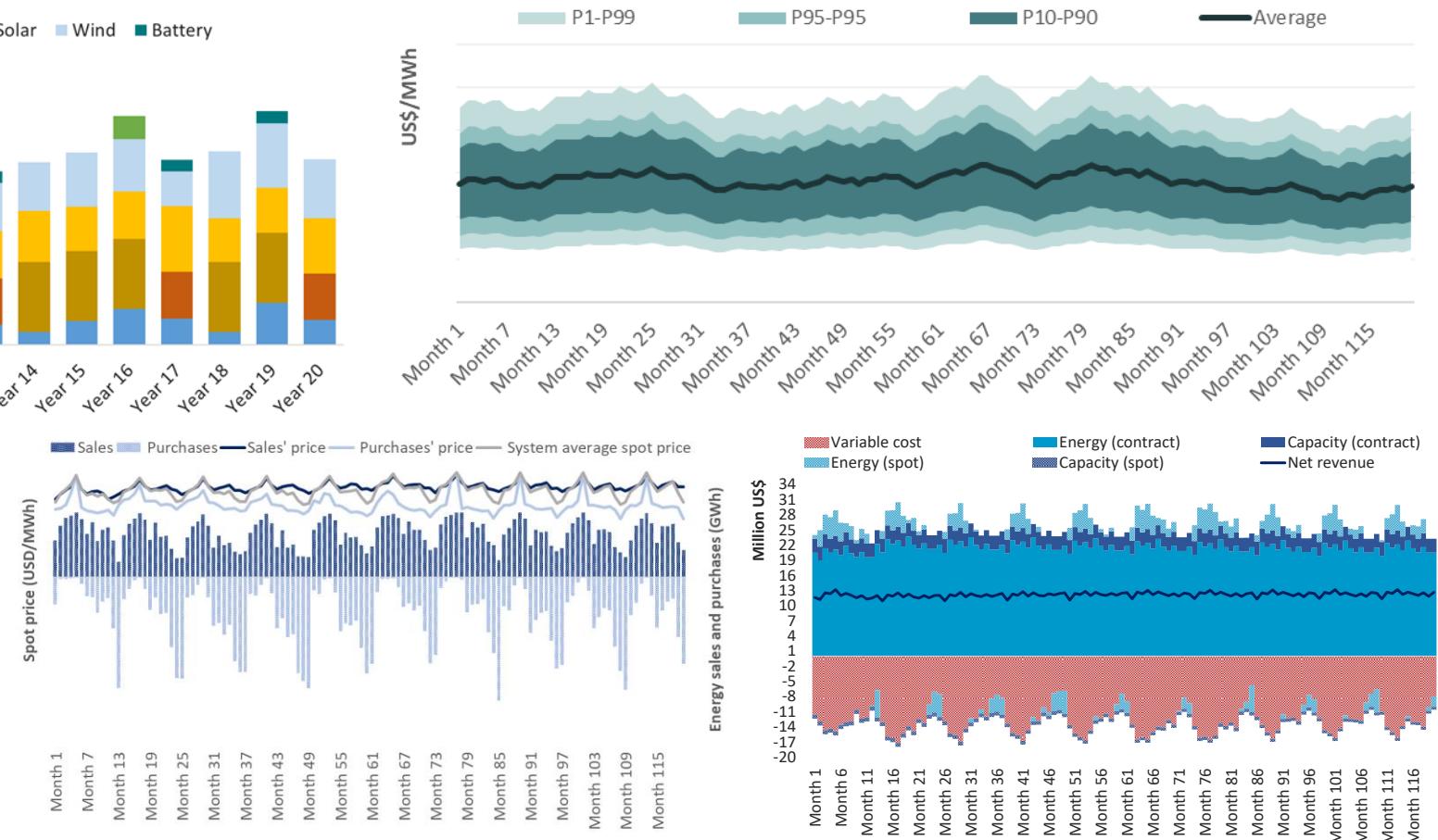
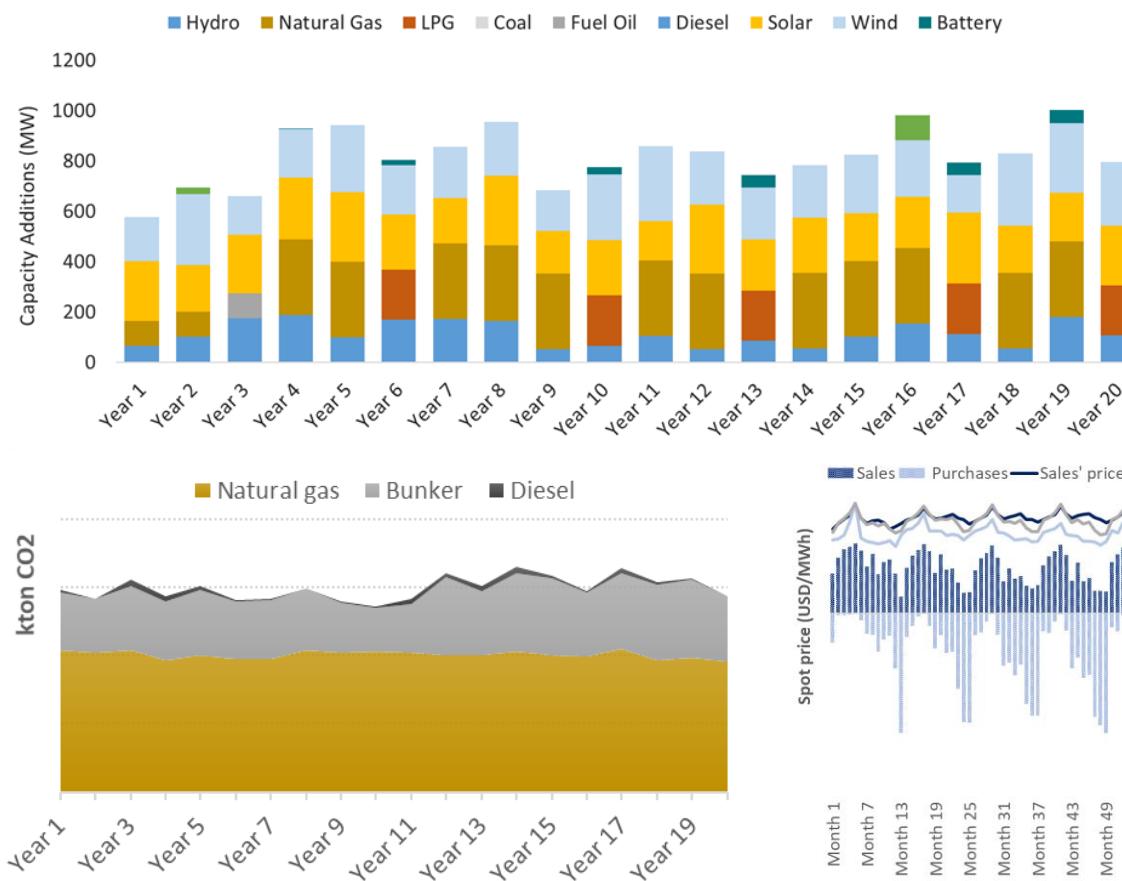


Minimize **Investment Cost + Operative Cost**

Subject to: **Energy Policy Constraints**

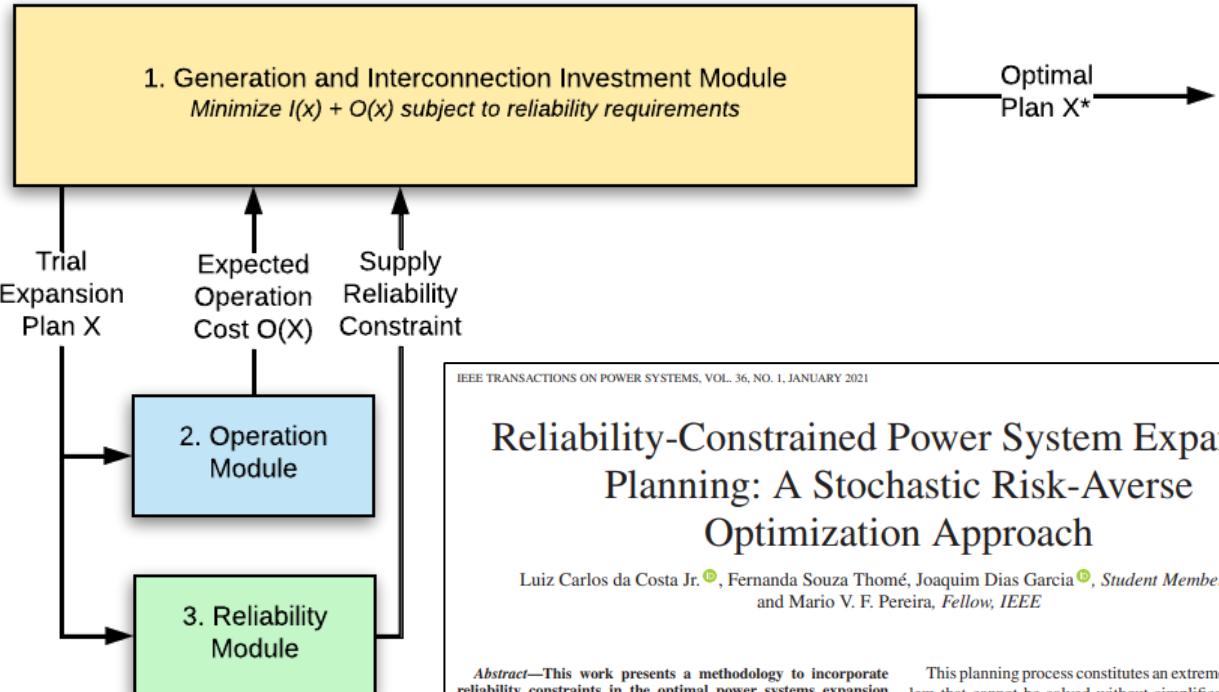
Example 1.0: Business as Usual

- Typical results to be evaluated are: system expansion, spot prices (with dispersion), emissions, generation, captured prices and revenues.



Planning 2.0: 1.0 + Reliability

- ▶ Security of supply under a wide range of conditions ("known unknowns")
- ▶ Addition of a reliability criterium (e.g., EPNS $\leq 1\%$)
- ▶ Supply reliability & resource adequacy
 - G&T forced outages
 - Composite generation and transmission reliability



IEEE TRANSACTIONS ON POWER SYSTEMS, VOL. 36, NO. 1, JANUARY 2021

Reliability-Constrained Power System Expansion Planning: A Stochastic Risk-Averse Optimization Approach

Luiz Carlos da Costa Jr. , Fernanda Souza Thomé, Joaquim Dias Garcia , Student Member, IEEE,
and Mario V. F. Pereira, Fellow, IEEE

Abstract—This work presents a methodology to incorporate reliability constraints in the optimal power systems expansion planning problem. Besides Loss Of Load Probability (LOLP) and Expected Power Not Supplied (EPNS), traditionally used in power systems, this work proposes the use of the risk measures VaR (Value-at-Risk) and CVaR (Conditional Value-at-Risk), widely used in financial markets. The explicit consideration of reliability constraints in the planning problem can be an extremely hard task and, to minimize computational effort, this work applies the Benders decomposition technique splitting the expansion planning problem into an investment problem and two subproblems to evaluate the system's operation cost and the reliability index. The operation subproblem is solved by Stochastic Dual Dynamic Programming (SDDP) and the reliability subproblem by Monte Carlo simulation. The proposed methodology is applied to the real problem of optimal expansion planning of the Bolivian power system.

Index Terms—System expansion planning, Benders decomposition, Power systems, Reliability, Stochastic programming, Risk measures, Mixed Integer Programming, Monte Carlo, Stochastic Dual Dynamic Programming.

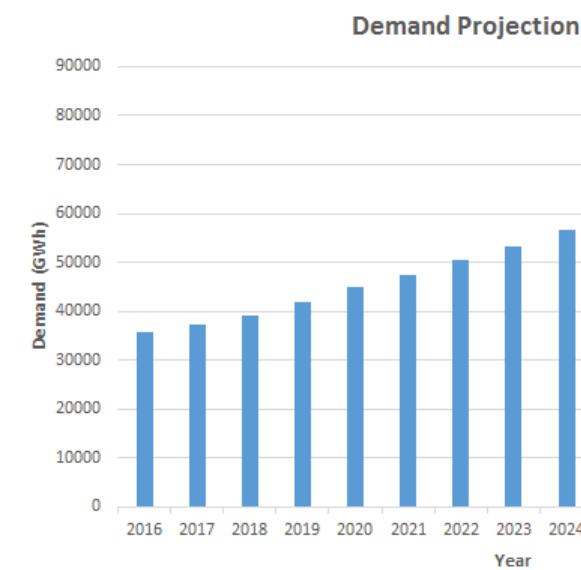
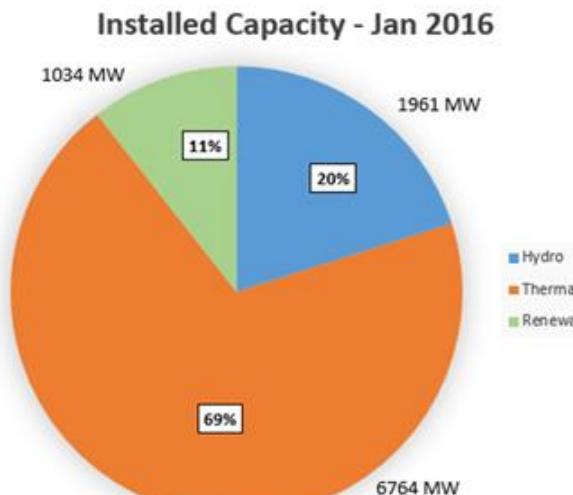
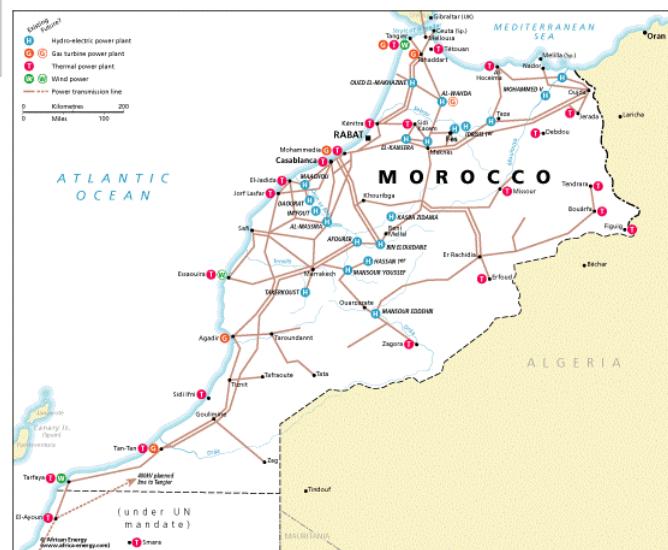
I. INTRODUCTION

THE goal of power system expansion planning (SEP) is to determine necessary changes in the system due to load growth, new technologies and policy-related constraints. In this

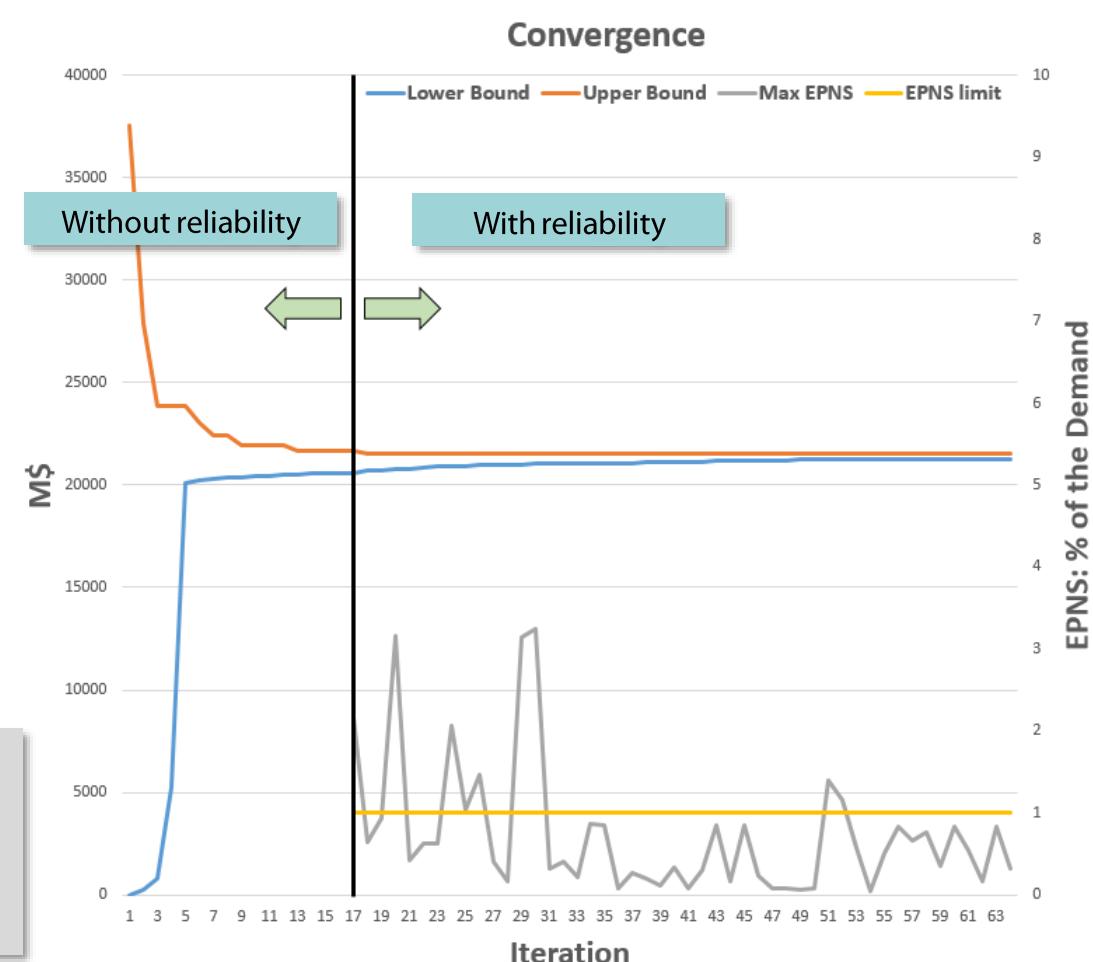
This planning process constitutes an extremely complex problem that cannot be solved without simplifications. Depending on the goals of the planner different aspects and details of power systems can be considered in general SEP. Many possible constraints are described in [7], some specific aspects include: carbon capture and storage [8]; unit commitment in the operation [9]; flexible demand and electric vehicles [10]; aggressive wind power penetration [11]. A common aspect in most SEP models is the representation of uncertainties, although each model typically focuses on specific sources of randomness, like renewable energy and load [12], outages or contingencies [13], [14]. Frequently used frameworks to deal with uncertainties in SEP are Stochastic Optimization [15], [16] and Robust Optimization [12], [17], both of which can also be combined [18].

Many techniques were proposed to solve the large-scale problems that arise from SEP modelling. Heuristics like Particle Swarm Optimization and GRASP were proposed in [8] and [19]. Many decomposition methods were presented due to natural scenario-wise and/or stage-wise structure: [20], [21] and [22] apply progressive hedging, Dantzig Wolfe decomposition was applied in [23] and Benders decomposition, perhaps the most used one, was applied in [12], [13], [15], [24]–[28].

Example 2.0: Moroccan expansion plan

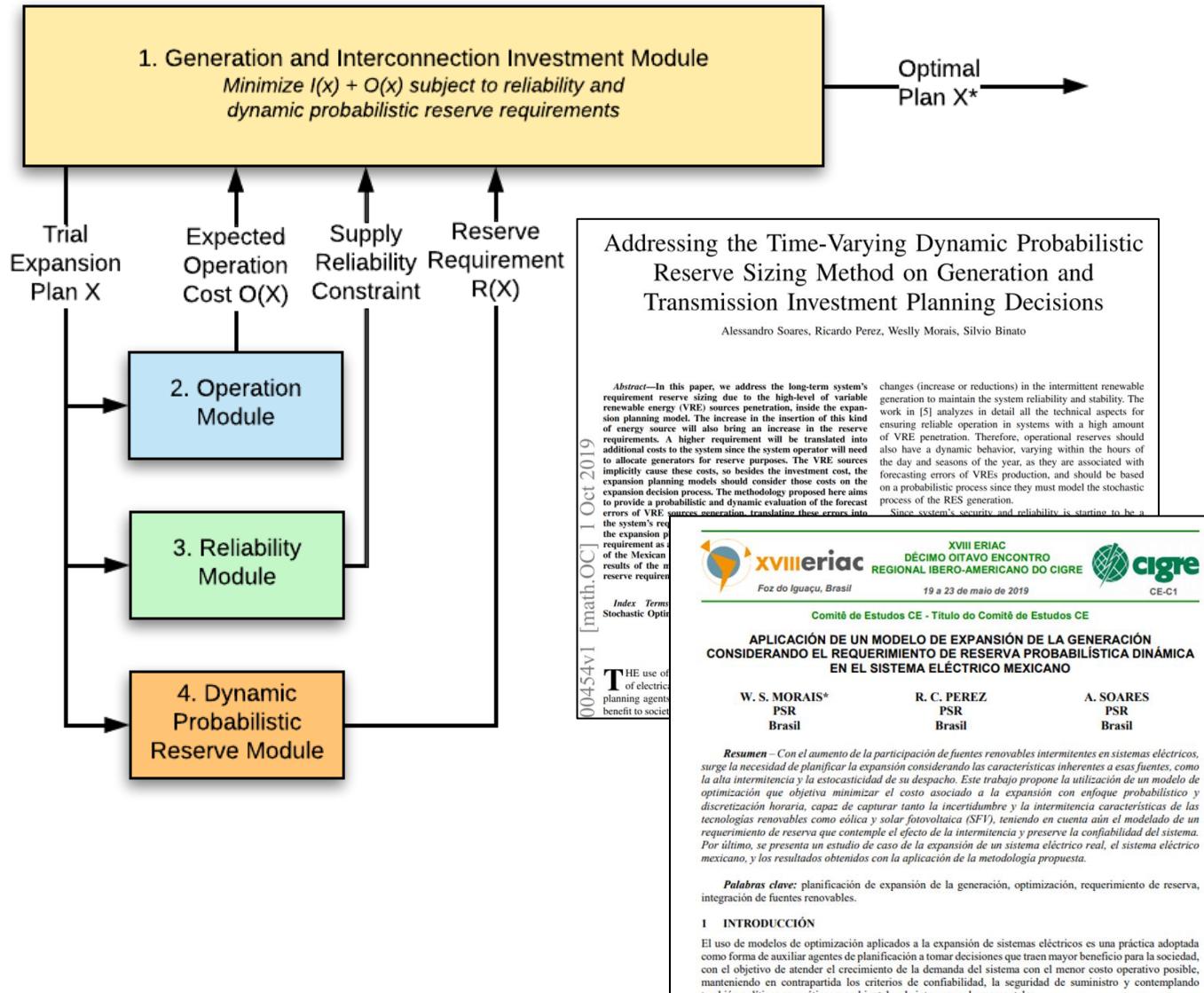


Planning horizon: 15 years
Yearly investment decisions
780 weekly operation stages (21 load blocks in each stage)

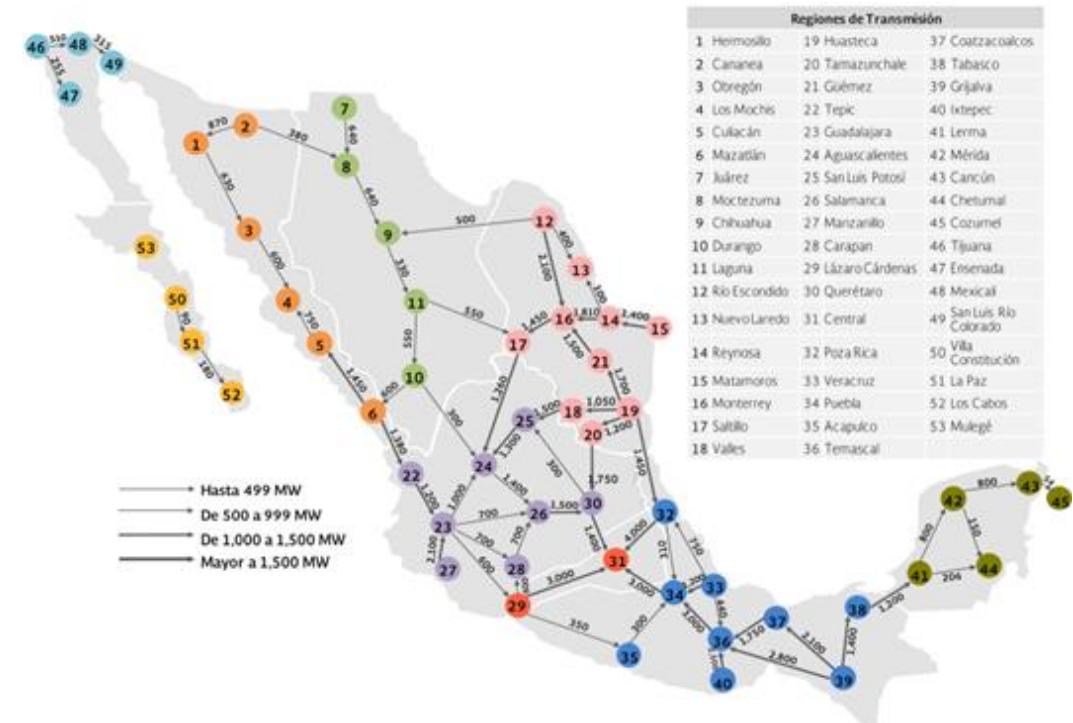
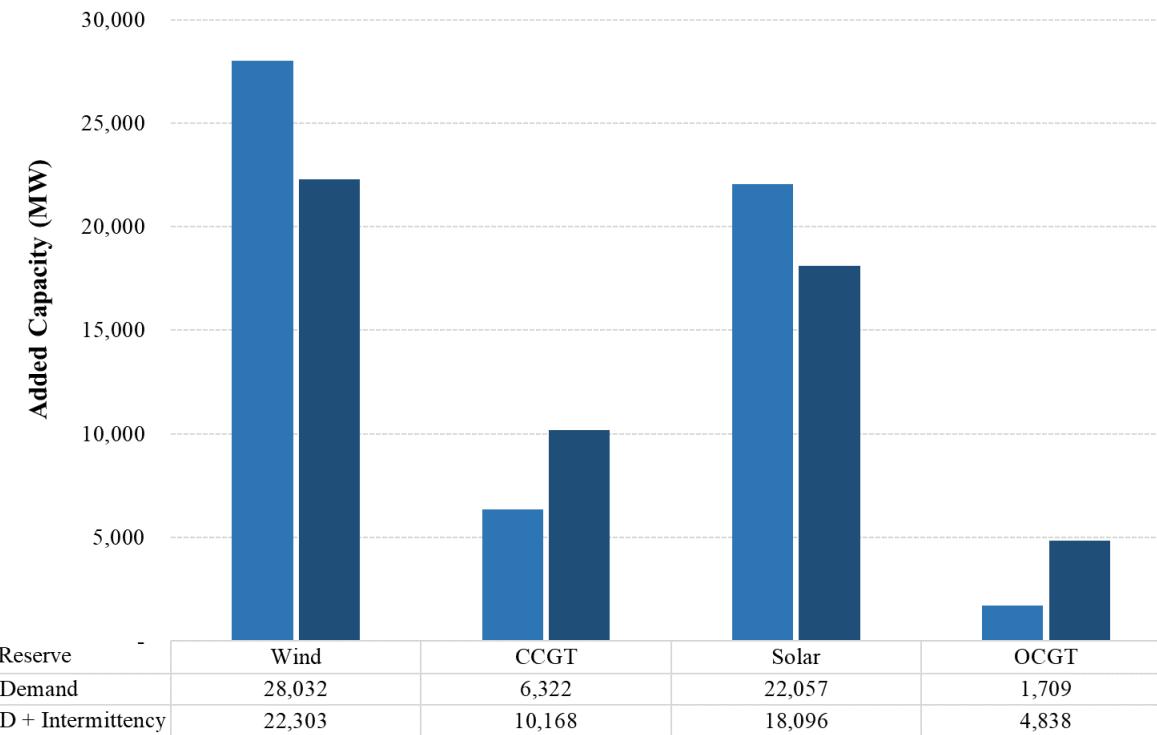


Planning 3.0: 2.0 + Co-optimization with Flexibility (Dynamic Probabilistic Reserve – DPR)

- The reserve requirement must be a decision variable for the expansion planning model
 - The DPR requirement will be co-optimized together with the investment decisions in an embedded way (endogenously)
- In other words, the model will select renewable plants considering the indirect cost that those plants will bring to the system by increasing the reserve requirements
 - The model may select, for example, specific renewable plants that, together, reduce the reserve requirement (Portfolio effect)
 - Wind parks that have a negative correlation coefficient (or high complementarity)
 - Solar plants that are not very close to each other (reducing the impact of the cloudy days)



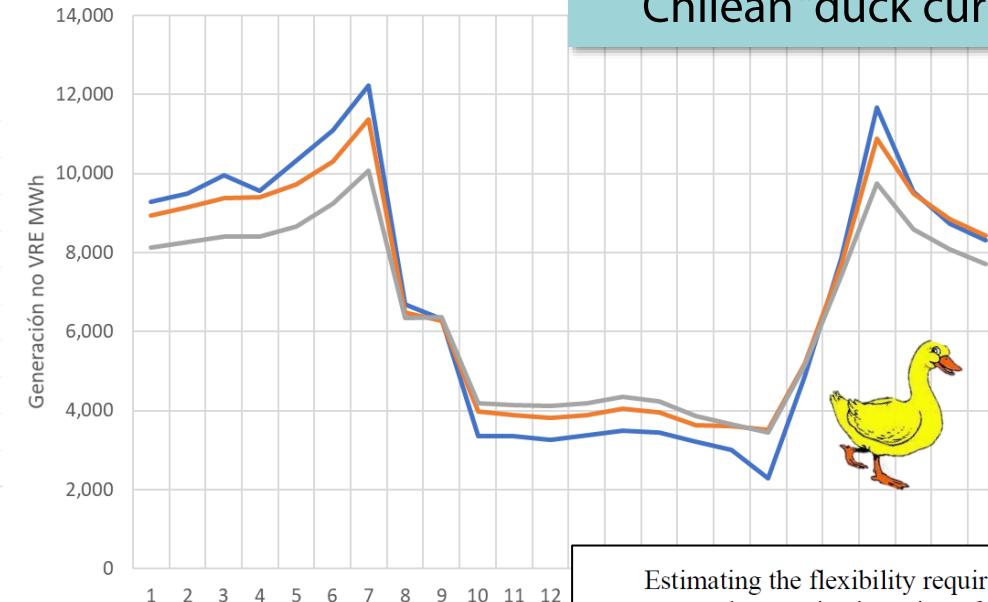
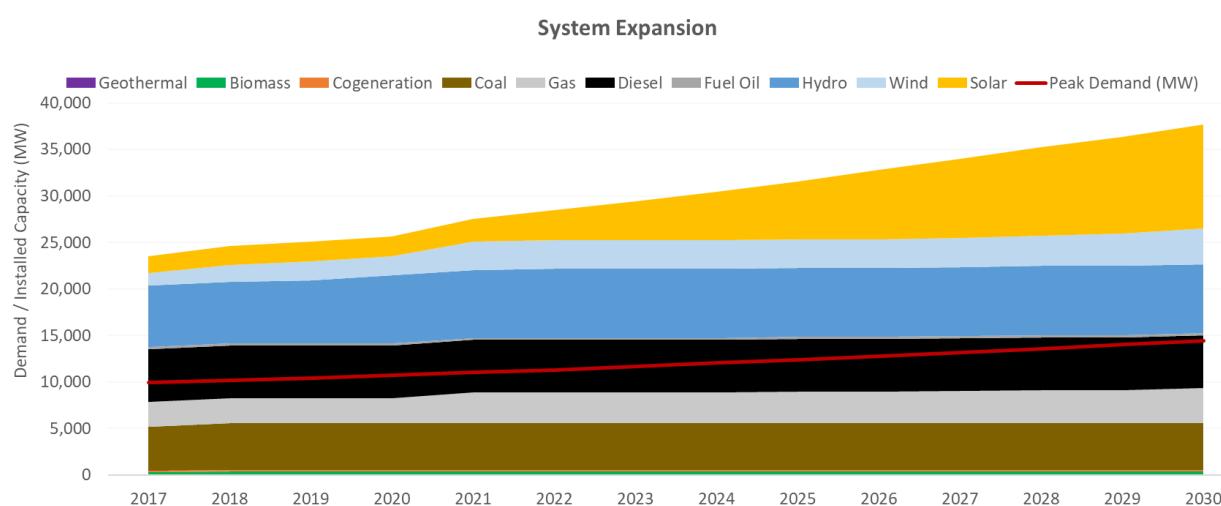
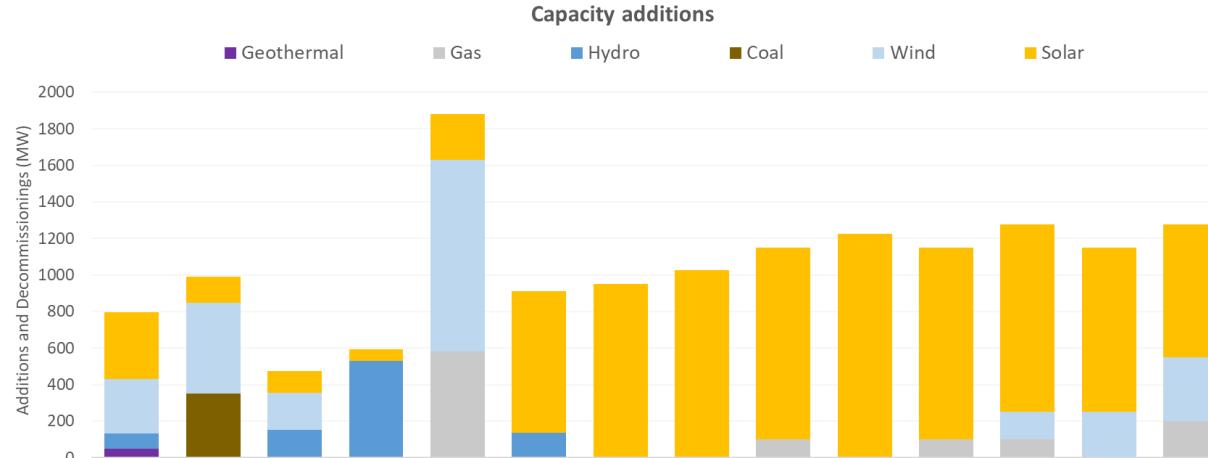
Example 3.1: Expansion of the Mexican System



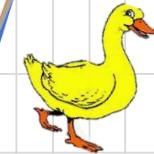
- The idea of the Dynamic Probabilistic Reserve (DPR) is to contemplate not only demand fluctuations and the failure of the biggest generating unit while calculating the 2ary Reserve Requirement, but also the **intermittency caused by renewables in an embedded way**
- When the DPR is considered in the investment model, VRE participation decreases due to an increased requirement of Secondary Reserve Requirement (increased operating costs)

Example 3.2: Chile

► The optimal expansion plan:



Chilean “duck curve” (2030)



Estimating the flexibility requirements and cost to manage the massive insertion of renewables in Chile

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Alessandro Soares and Silvio Binato
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Rodrigo Quinteros, Sébastien Mocarber,
Federico Heisig and Sébastien Rojas
Moray Energy Consulting
Santiago, Chile
Contact: quinteros@morayenergy.com

Abstract—This paper describes the methodology, computational tools and results of a study sponsored by the Chilean Generators’ Association (AGC) to quantify the investment and direct/indirect costs of providing the flexibility services probabilistic dynamic reserve, modulation etc.) required for the efficient and secure operation of the Chilean power system under the projected massive variable renewable energy (VRE) insertion in the next decade.

Index Terms—Flexibility service, Operational reserve, Optimization, Power systems, Variable energy resources.

I. INTRODUCTION

Chile’s variable renewable energy (VRE) resources are among the world’s competitive; the Atacama Desert has ideal insolation conditions to generate solar energy, produced higher in the nighttime, thus complementing the daily solar profile. Therefore, solar and wind won most of the contract auctions in the past years and should dominate the country’s expansion. The massive VRE insertion motivated government and agents to assess the *flexible generation reserve* required to manage their production variability: amount (type e.g. existing hydro and new gas generation), planned storage batteries etc.) and cost (investment cost or net cost, i.e. additional O&M costs due to increased cycling, missed energy sales of generation allocated to reserve etc.).

the *flexibility services cost* (investment, additional O&M costs) based on detailed probabilistic hourly simulations of system operation.

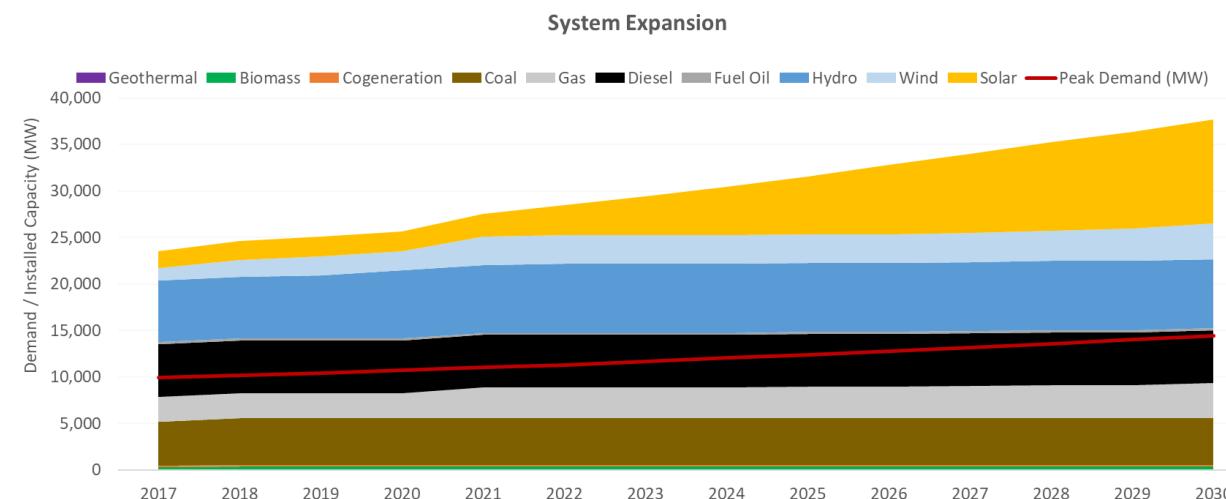
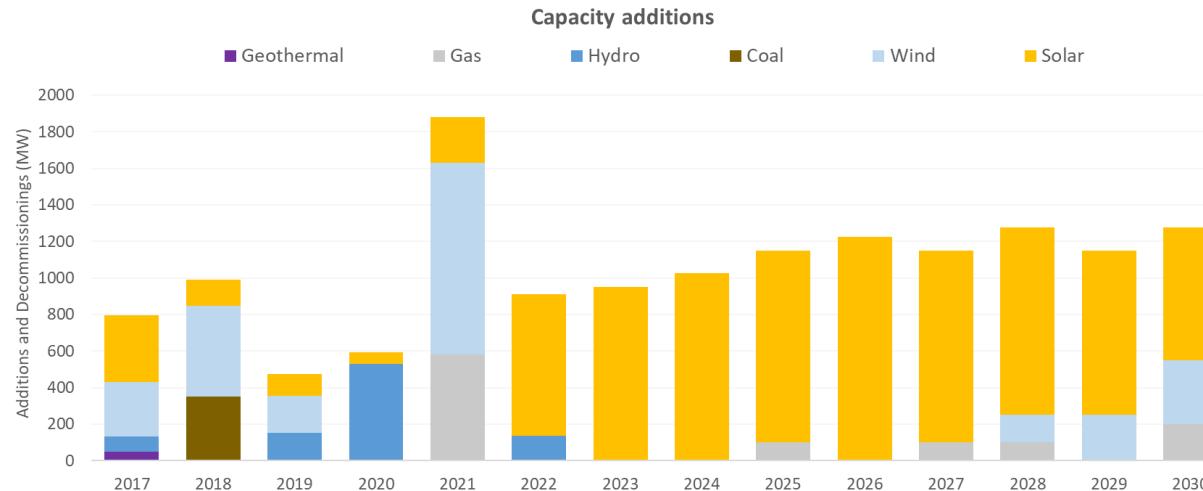
A. Related work

Flexibility is defined as the capability of a system to deal with unexpected changes in generation or demand while keeping a certain level of reliability at minimum cost [1][2]. Therefore, it is an attribute of great interest when integrating a significant share of variable sources [6]. IEEE PES’s study in [7] identifies the costs associated with flexibility services for thermal plants, such as increase on number of start-ups/cycling of generating. In [8], the start and stop costs for hydro generators are evaluated; the study in [9] quantifies the costs for German power plants considering up to 50% VRE insertion and concludes that direct and indirect start-up costs are higher than ramping-related costs. Another German study [10] quantifies the number of start-ups from 2010 to 2030 and show that they are affected by changes in the generation mix, in particular, retirement of nuclear plants.

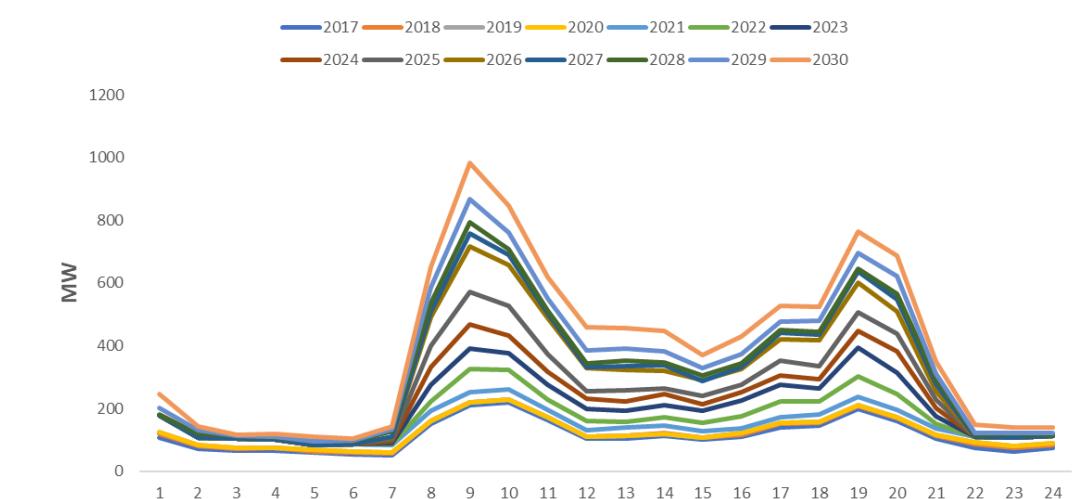
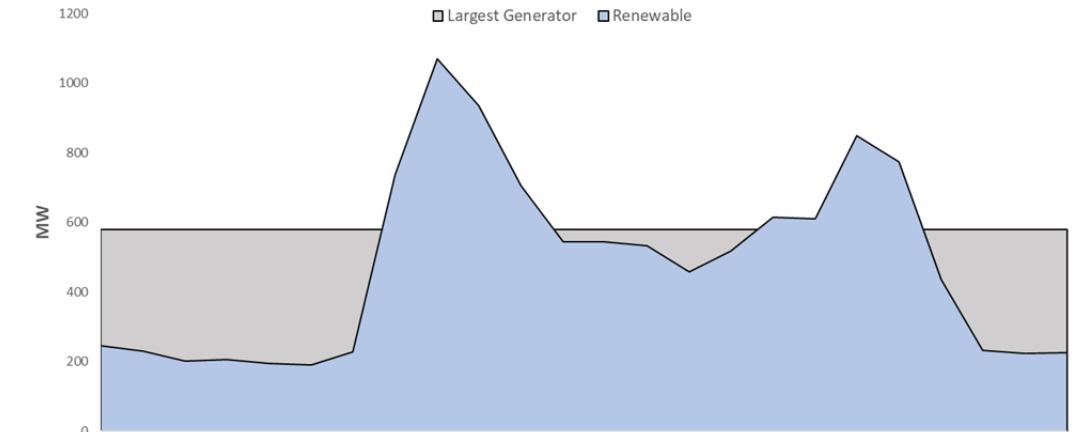
As stated in [3], just increasing the reserve margin (peak demand minus total installed capacity) is not enough to handle generation sources in case of VRE, in order to calculate the optimal mix of generation, reserve and flexibility, some short-term constraints such as upward/downward ramps, minimum up/down time and others must be considered. For

Example 3.2: Chile

► The optimal expansion plan:

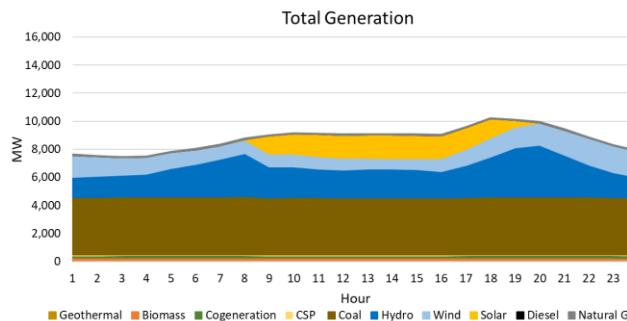


► Calculation of the Dynamic Probabilistic Reserves:

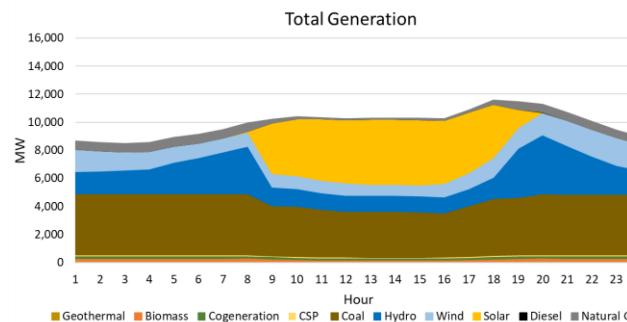


Example 3.2: Chile

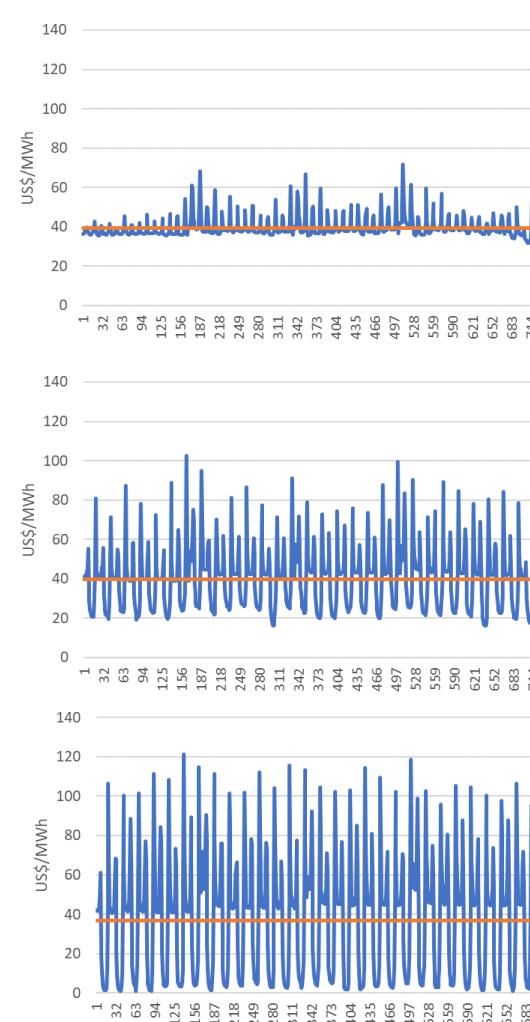
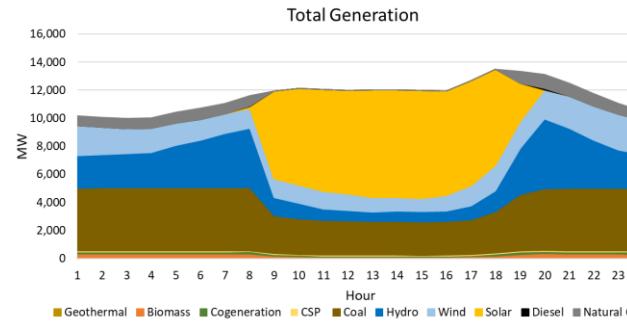
2021



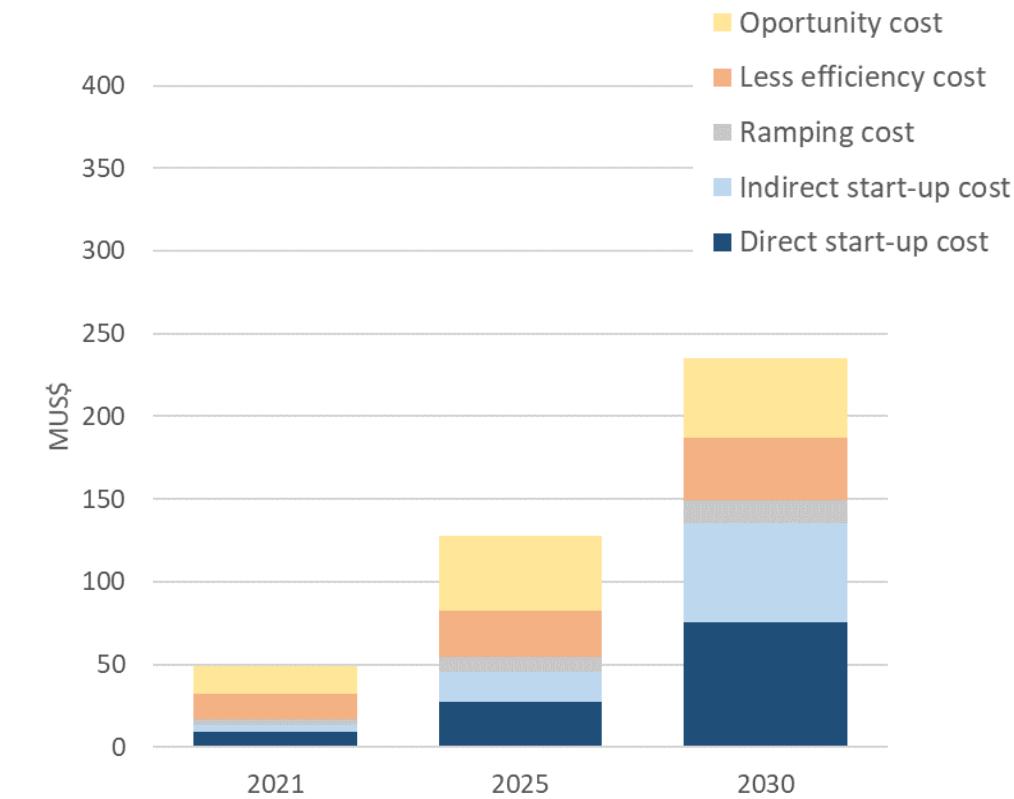
2025



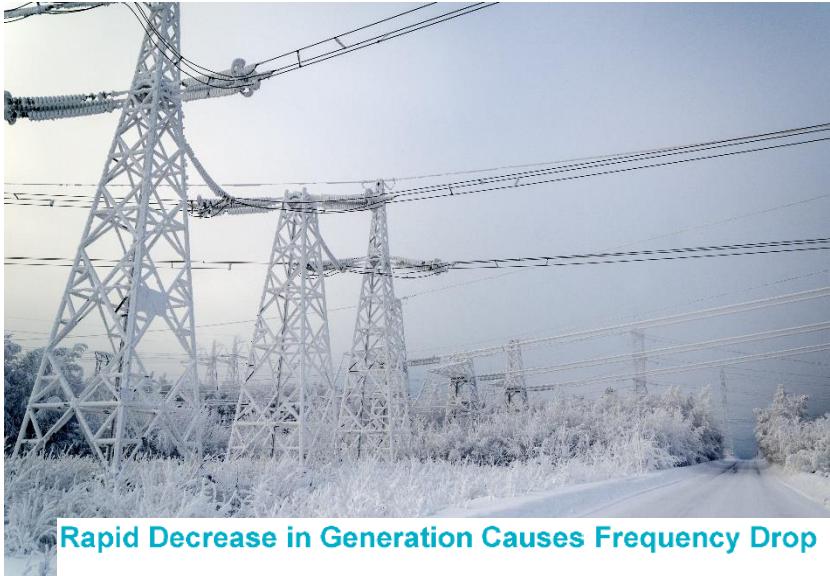
2030



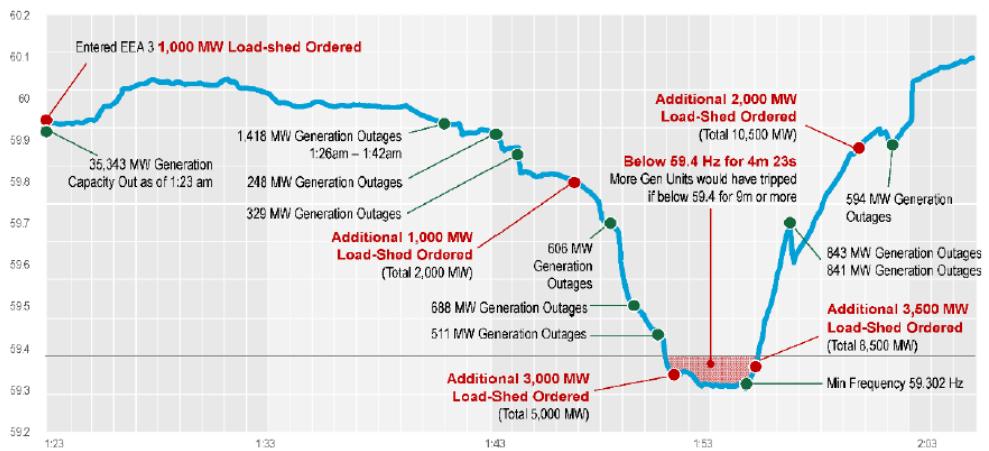
Example of the Flexibility Cost Calculation:



Looking back, what is the missing piece?



Rapid Decrease in Generation Causes Frequency Drop



Climate change is making India's brutal heat waves worse

As temperatures top 110°F, the heat could be deadly, especially to those without access to cooling.

By Casey Cawnhart

April 28, 2022

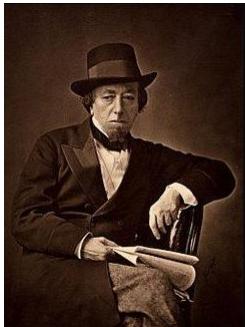


NEWS | 28 April 2022

Climate change will force new animal encounters – and boost viral outbreaks

Modelling study is first to project how global warming will increase virus swapping between species.

Looking back, what is the missing piece?



I am prepared for the worst, but hope for the best.

(Benjamin Disraeli)

PSR explica os tempos de resiliência na área energética
Uma discussão baseada no PSR Energy Report de abril/22

Rafael Kelman Fernanda Thomé

Dia 19 de maio, às 11

► ► ◀ 0:00 / 1:04:19
#EnergyReport #TransiçãoEnergética #MegaWebinar

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ENERGY REPORT April 2022 – issue 184

EFFICIENCY RELIABILITY FLEXIBILITY RESILIENCE

RESILIENCE: THE NEW ELEMENT OF ENERGY PLANNING

OPINION

The objective of this Editorial is to illustrate the incorporation of resilience in energy planning methodologies through two case studies with computational models developed by PSR: (i) an

2

Premium BR



Very hot topic at the IEEE PES GM 2022

2022 IEEE Power & Energy Society General Meeting
July 17 - 21, 2022 | Denver, Colorado

DENVER

PSOPE - Task Force on Advanced Intelligence Techniques for Resilient Power System Restoration
Monday, 18 July

Managing the Unexpected: Wildfire Risk in Power Infrastructure
Monday, 18 July

PSDP Task Force on Methods for Analysis and Quantification of Power System Resilience
Monday, 18 July

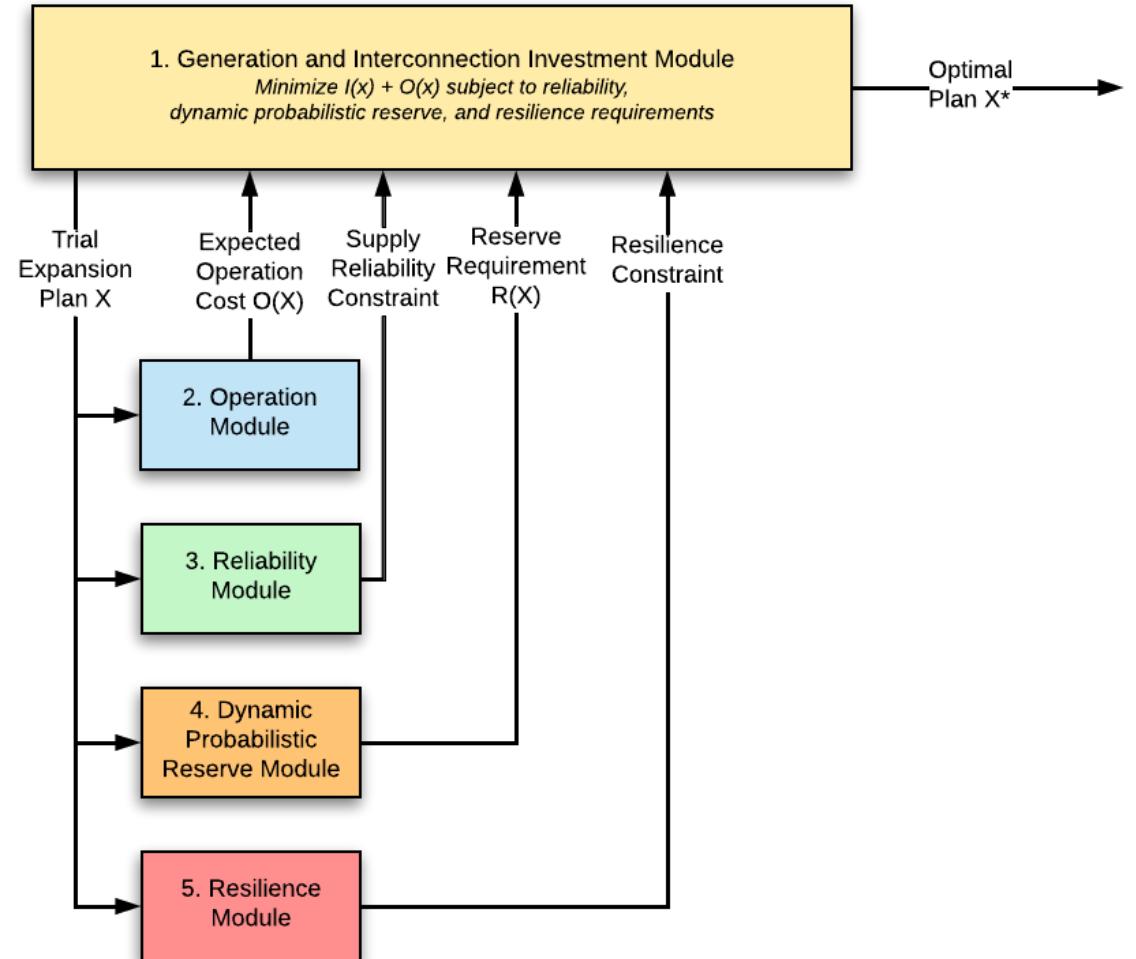
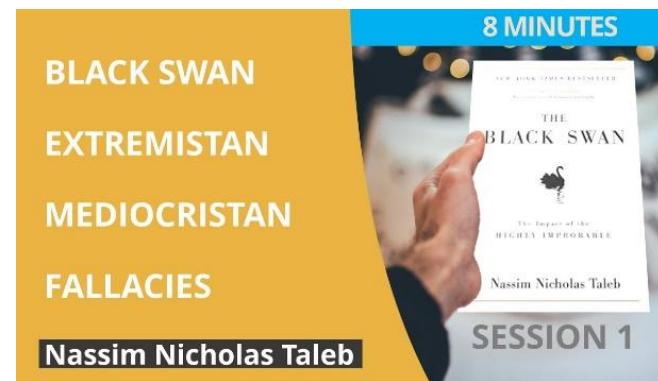
Future Reliability and Resilience Study of Electrical and Gas Systems under Extreme Weather Events
Monday, 18 July

15-17

17-19

Planning 4.0: 3.0 + Resilience

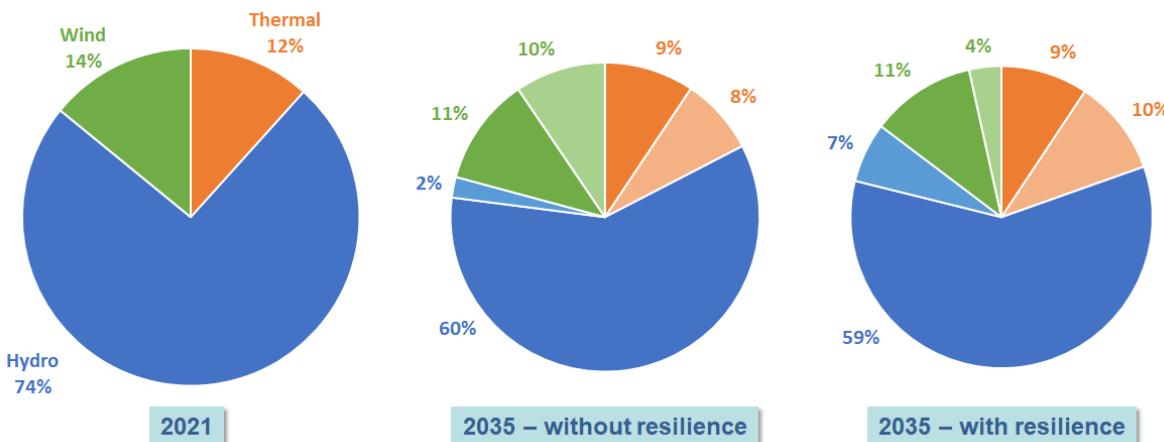
- ▶ Let's model now the “unknown unknowns”!
- ▶ Extreme **scenarios** (without assigning a specific probability):
 - Long-lasting decrease in renewable generation
 - Droughts and very severe temperatures due to climate change
 - Interruptions in fuel imports
 - Long-lasting outages of large generation trunks (e.g., due to cyber attacks, etc.)
- ▶ Resilience measures
 - Source diversification
 - Storage distribution
 - Greater self-sufficiency in each region



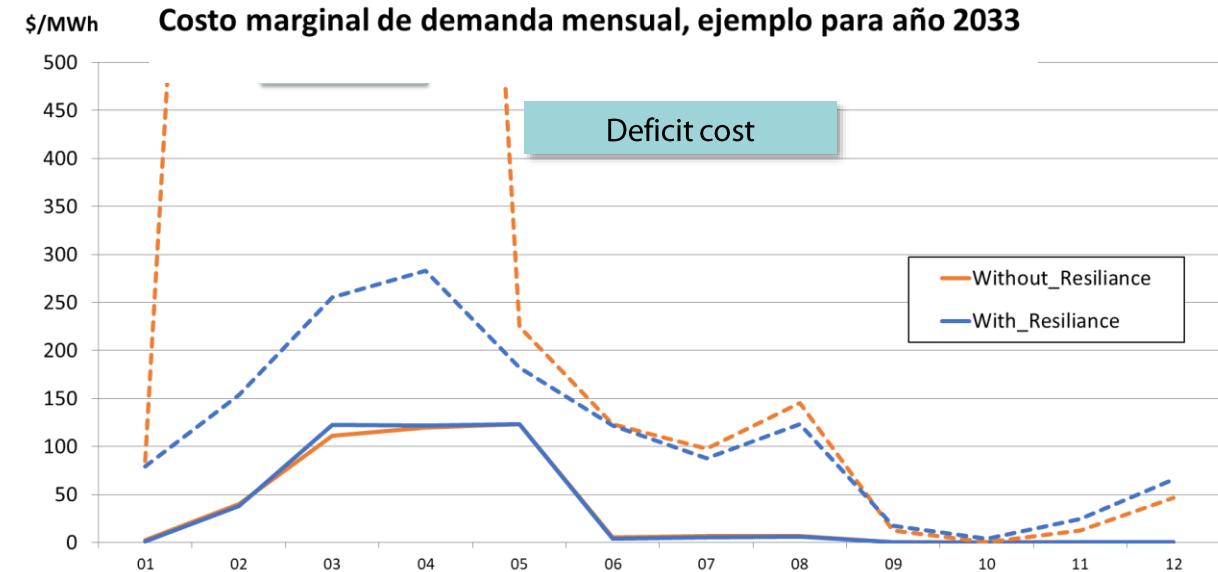
Captures high-impact low-probability events

Example 4.0: Costa Rica

- “Black swan” scenario is a severe reduction of wind



- Load Marginal Cost (2035)



Change in the mix: more diversification with less wind power

The critical event is faced with lower load marginal costs

*A single extreme scenario was considered, which is the almost total reduction of wind generation for one year. This scenario is totally artificial (we hope!) and was intended to highlight the effects of the resilience module

Summarizing everything!

► **Planning 1.0:** economic efficiency + energy policy guidelines

- BAU: Business as Usual
- Minimize expected total costs (investment + expected value of the operating costs) under uncertainty (variability of hydro, wind, solar, biomass production, and demand)

► **Planning 2.0:** (1.0) + reliability & resource adequacy

- Security of supply considering the “known unknowns” (equipment failure, composite generation and transmission reliability)

► **Planning 3.0:** (2.0) + co-optimization with flexibility (= dynamic probabilistic reserve)

- Contemplating: intermittency of renewables, demand fluctuations and failure of the largest generating unit
- Considers that reserve requirements vary over time → better dimensioned requirements and, consequently, the costs for the provision of these services are optimized

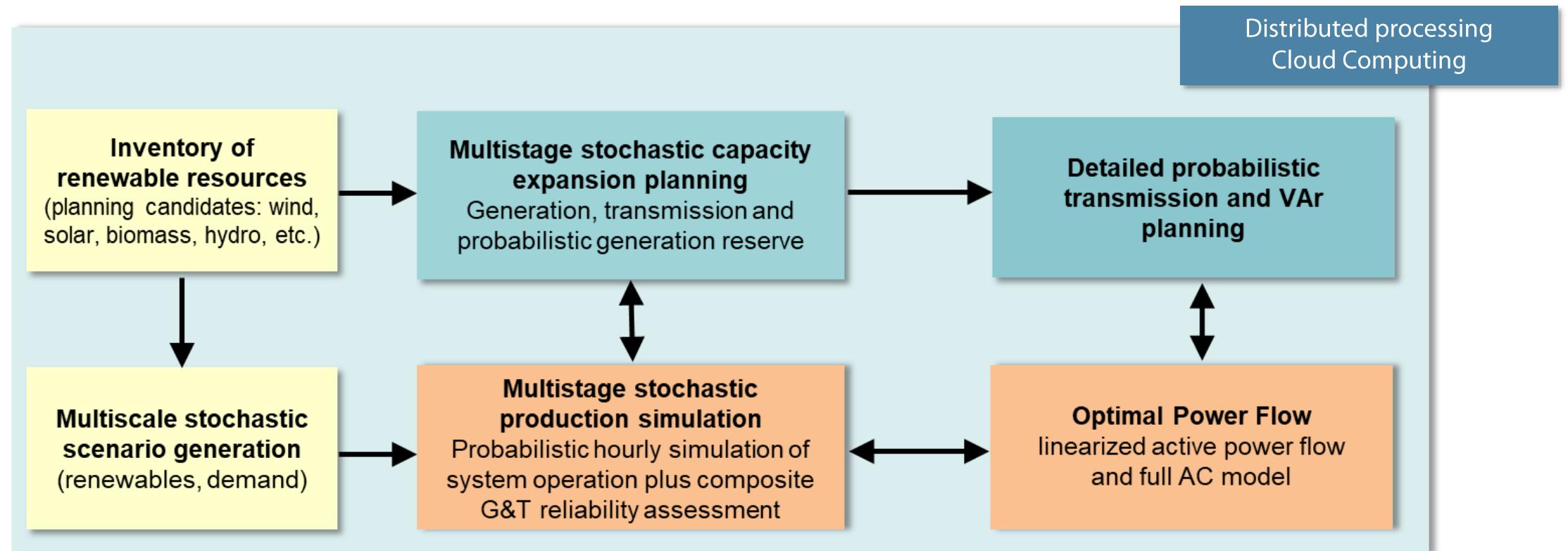
► **Planning 4.0:** (3.0) + resilience

- Feasible supply considering the “unknown unknowns” (stochastic models are not suitable for extreme events)
- Application examples: Geopolitical disruptions of fuel supply, very severe droughts, wind disruptions, effects of climate changes, etc.

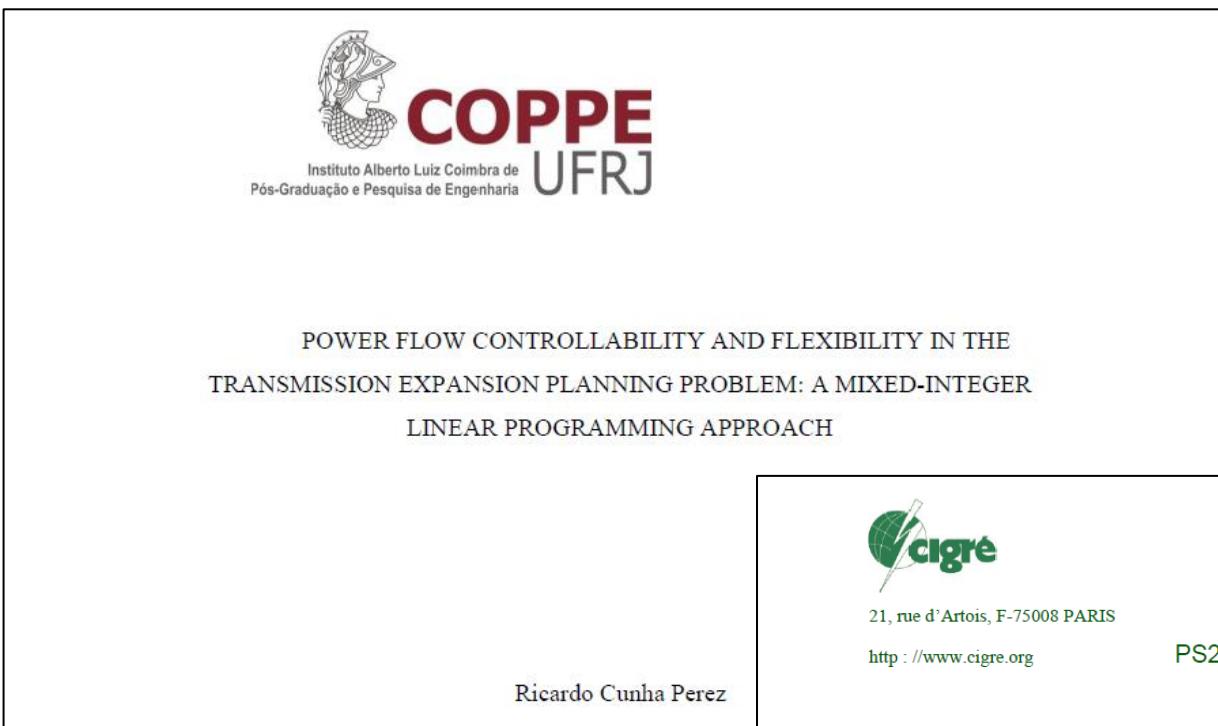
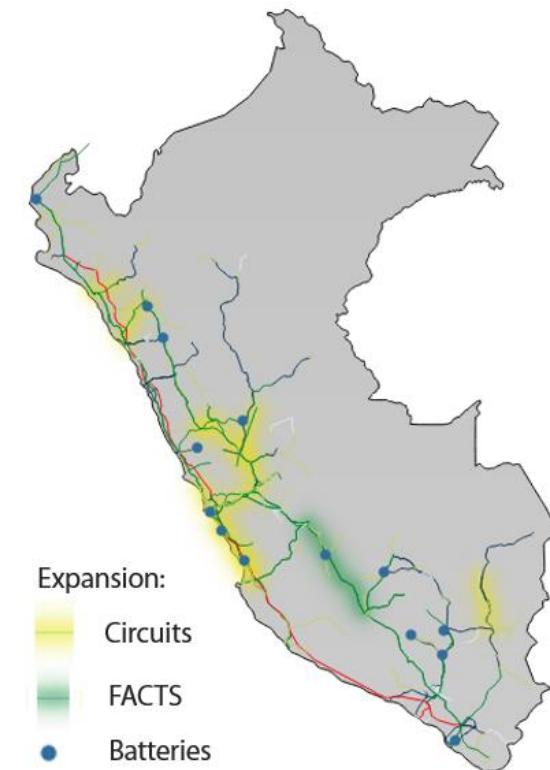
Integrated planning with “state of the art” tools is essential



Distributed processing
Cloud Computing



Out of the scope: flexibility in transmission



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Study Committee C1
PS2 – New System Solutions and Planning Techniques

CIGRE 2014

FACTS and D-FACTS: The Operational Flexibility Demanded by the Transmission Expansion Planning Task with Increasing RES

R. C. Perez*, G. C. Oliveira, M. V. Pereira, D. M. Falcão, F. Kreikebaum, S. M. Ramsay

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SUMMARY

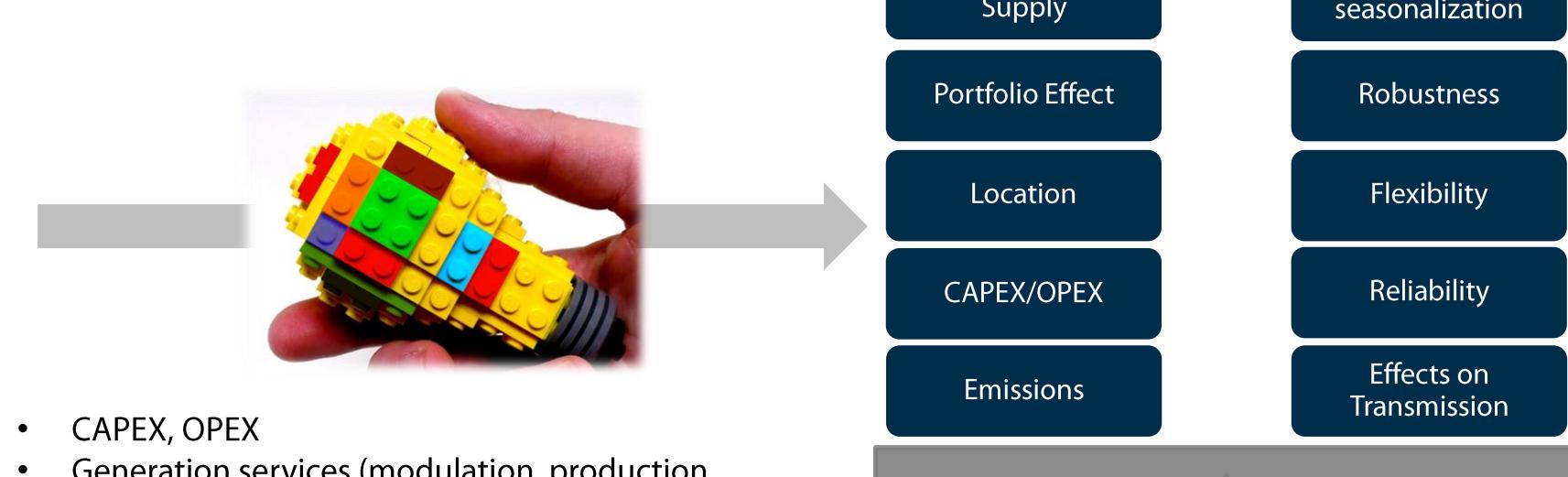
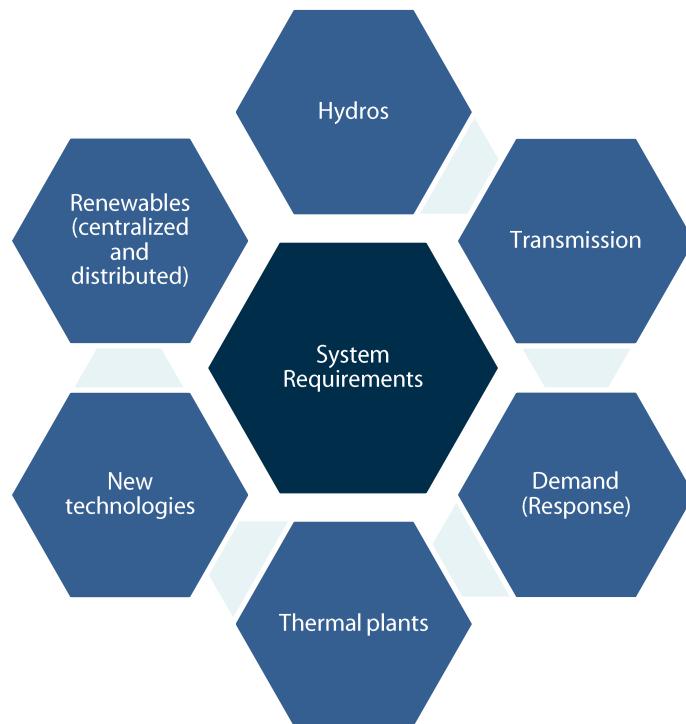
There are several reasons to explain why transmission system loading is less than 100%. These reasons include: (i) reliability, (ii) uncertainties associated with the demand growth forecast and (iii) different dispatch scenarios due to Renewable Energy Sources (RES), i.e., hydroelectricity, modern biomass, wind and solar power. The combination of these facts leads to high investments in the transmission systems to meet different dispatch scenarios and low loading throughout the year.



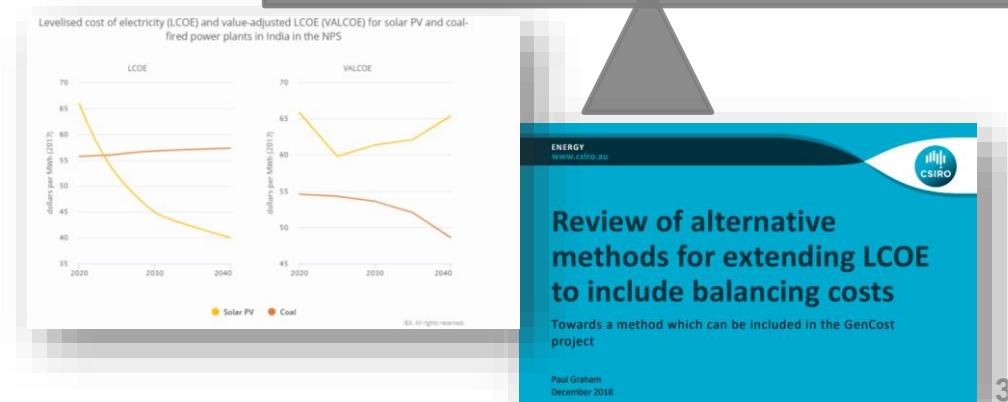
Conclusions

Valuation of the attributes of each technology

- Each technology offers a different value for the services that the system needs
- Each technology is a piece of the “puzzle” of the energy transition / decarbonization process
- The proposed planning scheme allows these values to be determined in order to compare them on the same basis, leveling the rules of the game



- CAPEX, OPEX
- Generation services (modulation, production synergies, flexibility)
- Infrastructure costs
- Effects on the transmission network
- Subsidies and incentives
- Environmental costs
- Emissions
- Resilience



In summary

- ▶ The complexity of power system planning & operation has increased substantially
- ▶ And market mechanisms are effective in promoting economic efficiency, **but they are not adequate to ensure the reliability of supply under a wide range of uncertainties**
- ▶ The reason is that reliability is a systemic attribute, that is, it requires an integrated vision of all supply and demand resources
- ▶ Recent disruptive events: economic decoupling of the United States and China; pandemic; extreme temperatures and flooding; and the invasion of Ukraine has also increased the importance of this integrated vision: in addition to reliability, it becomes necessary to ensure the resilience of energy systems
- ▶ **The importance of enhanced planning schemes is going to grow considerably**
- ▶ Fortunately, suitable computational tools are available in the market 

THE FUTURE OF OPTIMIZATION IN ENERGY IS VERY EXCITING, USEFUL AND FUN!



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Questions? Thanks!



PSR

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