### THE VALUE OF STOCHASTIC PROGRAMMING FOR ENERGY SYSTEMS PLANNING

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# **Research Team**



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# Introduction

- Energy system models should reflect the reality that planners must make decisions prior to the realization of future uncertainties
- Multi-stage stochastic linear programs (MSLP) optimize over future possibilities to yield a near-term decision strategy
- We use the expected value of perfect information (EVPI) and the value of the stochastic solution (VSS) as metrics in long-term capacity expansion of energy systems modeled as MSLP







# Overview

- Introduction
- Planning Capacity Expansion for Energy Systems
  - Background
  - Modeling and Characteristics
  - Tools for Energy Model Optimization and Assessment (TEMOA)
- Energy Systems Planning Problem as a Stochastic Program (SP)
  - ESPP in a multi-stage stochastic programming scheme
  - Uncertainty representation
  - Metrics to asses the value of SP for ESPPs
- Case Study
- Remarks and Comments

### Planning Capacity Expansion for Energy Systems

# Background

□ The energy system planning

requires a deep knowledge from the decision maker about **commodities**, **technologies**, **demand**, **costs** and network infrastructure



- Decisions in such systems are driven by information available at the present time as well as future projections
- To expand the existent installed supply-demand energy infrastructure, one has to consider the intrinsic characteristics of each subsystem (or sector) represented in the process

# Some Literature...

- Expansion planning of electricity power generation (Massé & Gibrat, 57; Jenkins & Joy, 74; Majumdar & Chattopadhyay, 99; Bistline, 15), transmission interconnections (Lee et al., 74; de la Torre et al., 08; Sauma & Oren, 07) and distribution networks (Asakura et al., 03; Hemmati et al., 15)



- Energy system planning, however, can be used in a much broader analysis including a combination of electricity, vehicle transportation, fuel supply-chains, district heating/coaling and other systems (Leung & Hsu, 84; Bhatt et al., 10; Chaudry et al., 14; Gironès et al., 15; Gómez et al., 16)
- The energy system planning problem (ESPP) is known for several decades (Kroneberg, 50) and one of the first formulations as a linear programming model is presented in (Massé and Gibrat, 57)
- Mathematical models have being extensively used to represent such problems (Jebaraj & Iniyan, 06; Connolly et al., 10), eg. Markal (Fishbone & Abilock, 81)

# Uncertainty and Metrics in MSLP



- Uncertainty plays a fundamental role in the definition of strong strategies which aim to minimize the total combined expansion and operational costs over a particular time horizon using mathematical models
- Understanding uncertainty is also a pre-requisite to correctly use such information in mathematical models (Kann & Weyant, 00)
- The representation of each uncertainty type will result in different models and different results and values that can be achieved with such models
- We aim to investigate long-term ESPP under uncertainty and what is the value of a MSLP representation of the problem instead of a deterministic
- Value of the stochastic solution (VSS) (Birge, 82)

How much money the hedging strategy saves relative to the total cost obtained by an optimization model when uncertainty is ignored

# Models and Characteristics

- Models for conducting energy system analysis:
  - Markal/Times
  - OSeMOSYS
  - Message





- Energy economy optimization model
- Technology assessment and policy analysis at ≠scales
- Model is implemented in a general algebraic formulation combined with
- Stochastic Programming capabilities (extensive LP and Progressive Hedging)

#### http://www.temoaproject.org https://github.com/TemoaProject/temoa/



# TEMOA – General Purpose



- TEMOA represents a capacity expansion and operational model for energy systems
- Represents a MSLP in a network with multiple technologies and multi-commodities
- Model's objective: minimize cost of energy supply over a defined time horizon (present + expected future cost)
- Processes represented in a macro level





Commodity flow balance

# **Mathematical Formulation**

$$\begin{split} \hline & \text{LeanCost} = \sum_{v,v} \left( \begin{bmatrix} c_{v,v} \cdot A_{v,v} \cdot \sum_{j=0}^{u} \frac{1}{(1+\text{CDR})^{v,v-k_j}} \end{bmatrix} \cdot \mathbf{CAP}_{v,v} \right) \\ \hline & \text{FixedCost} = \sum_{j,v,v} \left( \begin{bmatrix} c_{r,u,v} \cdot \sum_{j=0}^{u} \frac{1}{(1+\text{CDR})^{v,p-k_j}} \end{bmatrix} \cdot \mathbf{CAP}_{v,v} \right) \\ \hline & \text{min} \\ & \text{ACT}_{FI,FO} \\ & \text{CAP,CAPVAL} \\ \hline & \text{s.t.} \quad \mathbf{ACT}_{p,s,d,t,v} = \sum_{i,o} \mathbf{FO}_{p,s,d,i,t,v,o} \quad \forall \{p, s, d, t, v\} \in \Theta_{activity} \\ & \text{Process} \\ & \text{activity} \\ \hline & \left( CF_{s,d,t,v} \cdot C2A_t \cdot SEG_{s,d} \cdot TLF_{p,t,v} \right) \cdot \mathbf{CAP}_{t,v} \geq \mathbf{ACT}_{p,s,d,t,v} \quad \forall \{p, s, d, t, v\} \in \Theta_{activity} \\ & \sum_{i,t,v} \mathbf{FO}_{p,s,d,i,t,v,c} \geq DEM_{p,c} \cdot DSD_{s,d,c} \quad \forall \{p, s, d, i, t \in T-T^s, v, o\} \in \Theta_{flow} \\ \hline & \text{Fo}_{p,s,d,i,t,v,o} \leq EFF_{i,t,v,o} \cdot \mathbf{FI}_{p,s,d,i,t,v,o} \quad \forall \{p, s, d, i, t \in T-T^s, v, o\} \in \Theta_{balance} \\ \hline & \text{Global} \\ & \text{commodity} \\ \hline & \text{Other constraints and bounds: baseload, emissions, battery storage,...} \end{split}$$

#### Energy Systems Planning Problem as a Stochastic Program

# ESPP as a MSLP



# Uncertainty Representation

- Uncertainties may be represented by scenarios of:
  - Economic grow
  - Commodity/technology price trajectories (Van der Weijde & Hobbs, 12)
  - Demand realization (Pineda and Morales, 2016)
  - Technology reliability (Hajipour et al., 2015)
  - Policies related to greenhouse gases emissions (Bistline & Weyant, 13; Park & Baldick, 15)
  - Renewable generation penetration (Munoz et al., 14)
  - Renewable resources availability (Gil et al., 15)
  - Technological, economic, and policy-related (Bistline, 15)

### **EVPI & VSS Computation**



### What Influences Decision-making in ESPPs?

#### Cost vectors

- Existent capacity
- Commodities demand
- Bottlenecks in the network
- Efficiency of processes
- Reliability
- Availability of resources
- Time-delay between decisions and physical use
- Linking decisions and constraints
- Bounds that will limit investments

#### Discount rate



# Value of Money & Time

□ Branch weighting (discount rate of 5% per 5 years)



### Case Study

# System Description

- How to design a power generation expansion plan for South Sudan?
- Significant size and abundant natural resources
- The country has
  ~ 30 [MW] of
  existent capacity



- Existent studies for hydropower investments
- However, is it the best option?





# **Problem Characteristics**





Solar investment cost: 3.35 \$/kWMinimize total costHydro: 2.5 - 13.2 \$/kW, Thermal: 1.5\$/kWMinimize total costThermal variable cost: 40 - 46 \$/MWhMost information extracted from:<br/>(Hatch Report, 2014)Curtailment cost: 300, 600 and 5000 \$/MWh(Hatch Report, 2014)Transmission lines investment cost, fixed costs, capacity factors, efficiency, etc

# **Problem Structure and Uncertainty**





	Conflict Scenario Probabilities	
Conflict at	High	Low
Hydro	0.45	0.05
Thermal	0.25	0.05
Both	0.25	0.05
No conflict	0.05	0.85

### **EVPI & VSS for Different Curtailment Costs**



# Other Problem Instances

- We keep the instance with high probability of conflict
- Consider two annual discount rates (10% and 5%)
- Analyze problem with od fferen and humber 20 and 1 stages



### EVPI & VSS (Curtailment, # of Stages and Discount Rate)



### Remarks & Comments

# Remarks & Comments

- We presented an overview about the ESPP and TEMOA
- We used EVPI and VSS as metrics to assess the value of representing this problem as a MSLP
- Problem's characteristics determines the optimization design (What to consider as uncertainty? How many time stages? Discount rate? Relationships with decisions from previous stages, etc)
- We briefly talked about a case study for South Sudan (explained in detail at the poster Section today at 12:30 PM by Neha Patankar)
- Future work will analyze EVPI and VSS considering other variations in model formulation as well as uncertainties

# Thank You !





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