# Selecting and Initializing Representative Days for Generation and Transmission Expansion Planning with High Shares of Renewables



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#### Outline

### Generation and Transmission Expansion Planning

Problem description

### Analysis Performed

- Selection of representative days
- Initialization of representative days

### Tests

Testing Framework

## Future Work







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Generation & Transmission Expansion Planning

#### **Generation and Transmission Expansion Problem**

- Determine a least-cost investment schedule for:
  - Construction of new generating capacity;
  - Building of new electrical interconnections;
  - Decommissioning of generating units.



- Given projections of the energy system evolution, define expansion plans in order to:
  - Supply load
  - Achieve policy goals

at **minimum** total cost (investment and operation).



- **Long-term** planning horizon;
- Hourly resolution needed to catch
  - Fluctuation of solar and wind power generation
  - Technical constraints on thermal power production
    - ✓ Minimum Up Time
    - Minimum Down Time
  - Dynamics of storage facilities
    - ✓ Hydro pumped storage
    - ✓ Battery storage
- High uncertainty.



Computationally intractable problem.

## **Analysis** Performed

- In order to keep the problem computationally tractable frequently a small number of representative days is considered.
- Different approaches have been proposed in the **literature**:

#### > Simple heuristics

Only days with minimum load, maximum load, largest daily load spread are considered.

#### > Clustering algorithms

Days with similar load, wind/solar production are grouped into clusters, with the cluster's centroid then taken as the representative day.

#### Load duration curve

Minimization of the difference between the load duration curve and the one reproduced by the representative days.



- Consider the **first year** of the planning horizon;
- Select days with **minimum** and **maximum** total load in the power system and remove them from the dataset.
- For each day *d* of the new dataset a vector *V<sub>d</sub>* is created which contains for all zones *z* the values of
  - ✓ Hourly load  $(D_{z,t}^d, z \in \mathbb{Z}, 1 \le t \le 24)$
  - ✓ Hourly solar capacity factor  $(\mu_{z,t}^d, z \in \mathbb{Z}, 1 \le t \le 24)$
  - ✓ Hourly wind capacity factor  $(\rho_{z,t}^d, z \in \mathbb{Z}, 1 \le t \le 24)$

In this way **correlations** within a day between load and renewable capacity factors in different hours of day and different system zones are taken into account.

- Normalize vectors  $V_d$ ;
- Define a **threshold** for the choice of the number of representative days (e.g. 1%).

a) By the k-medoids algorithm compute k clusters so as to minimize the deviation between vectors  $V_d$  and their representative  $V_c^*$ :

$$\min \sum_{c} \sum_{d \in D_{c}} \|V_{d} - V_{c}^{*}\|^{2}$$
$$D_{c}: \text{ group of days } d \text{ in cluster } c$$

- b) Associate to each representative day  $V_c^*$  the weight  $|D_c|$ , i.e., the number of historical days grouped in cluster c
- c) Construct the load duration curve corresponding to the representative days and compute its distance to the original load duration curve
- d) If the mean absolute percentage error in the load duration curve approximation is lower than the input threshold, stop; otherwise increase *k* by one and repeat.



- The k + 2 representative days for the first year of the planning horizon have been determined:
  - $\succ$  k centroids identified by the k-medoids algorithm;
  - 2 extreme days.

Determine the representative days of the subsequent years by applying annual growth factors to load profiles.

• Use representative days to evaluate power system **operation** with hourly resolution in the expansion planning model.



- The use of representative days raises the crucial issue regarding how these days should be **linked** in the expansion planning model.
- Most of the existing methods consider the representative days as temporally consecutive, linking these days according to an arbitrary order
  - The order chosen could affect the model results;
  - The interconnection among days increases computational costs and prevents from exploiting the decomposable structure of the expansion planning problem.
- In our method, we assign to each thermal power plant an **initial** ON/OFF **status** in every representative day by means of a decision tree built on historical data.





- We consider a **training set** including commitment decisions for Italian thermal power plants during a year.
- We build parameters  $\gamma_{k_0}^d$ , which describe the ON/OFF status of thermal power plant k in the last hour of day d 1,  $2 \le d \le 365$ .
- We compute the following features:
  - Marginal cost ratio;
  - Start-up cost;
  - Minimum up time;
  - Minimum down time.
- We estimate on the training set a **decision tree**, in order to identify a classification rule that could determine the initial ON/OFF status according to features values.





- The decision tree is used to assign to each thermal power plant k in every year y of the planning horizon the **probability**  $\pi_k^y$  of having an initial ON status.
  - ✓ Thermal production costs change throughout the planning horizon, thus thermal power plants may present different probabilities  $\pi_k^{\gamma}$  along the planning horizon.
- Parameters  $\pi_k^{\mathcal{Y}}$  are used to **set** the probability of extracting 1 in the random selection between 0 (i.e., OFF) and 1 (i.e., ON).
- For each thermal plant k and for every year y, this random selection is **repeated** for all representative days, in order to assign to each representative day  $c \in C^y$  a specific initial status  $\gamma_{k_0}^c$ .



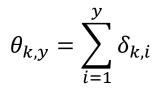


- A simplified model for GTEP is used as a **testing framework** to assess the performances of the proposed method:
  - System cost minimization formulation;
  - > Only thermal, wind and solar power technologies are considered;
  - Zonal representation of power system;
  - Transportation model for power exchanges among zones;
  - Inelastic demand.

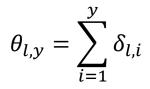




- For each **year** *y* of the planning horizon
  - $\delta_{k,y}$  Building of new thermal power plant k
  - $\theta_{k,y}$  Availability of new thermal power plant k:



- $\delta_{l,y}$  Building of new transmission line l
- $\theta_{l,y}$  Availability of new transmission line l:



- $sol_{z,y}$  New solar installed capacity in zone z
- $wind_{z,y}$  New wind installed capacity in zone z



- For each **hour** *t* of every **representative day** *c* 
  - $p_{k,t}^c$  Power output unit k above the minimum
  - $\gamma_{k,t}^{c}$  1: unit k is ON; 0: otherwise
  - $\alpha_{k,t}^{c}$  1: unit k is started-up; 0: otherwise
  - $\beta_{k,t}^c$  1: unit k is shut down; 0: otherwise
  - $x_{l,t}^c$  Energy flow on transmission line l
  - $RES_{z,t}^{c}$  Renewable generation in zone z
  - $ENP_{z,t}^{c}$  Energy not provided in zone z
    - $OG_{z,t}^{c}$  Overgeneration in zone z



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#### **Objective Function**

$$\min z = \sum_{y \in \mathcal{Y}} \frac{1}{(1+r)^{y-y_0}} \begin{bmatrix} \sum_{k \in \mathcal{K}_C} IC_k^{th} \ \delta_{k,y} & \text{Investment content of thermal power of the thermal power of thermal power of the thermal power of the the thermal po$$

ent cost for new power plants

ent cost for new RES

ent cost for new ssion lines

production cost

CES



#### Constraints

- Load Supply
- Reserve Requirements
- Renewables Penetration
- Energy flows on transmission lines
- Operation of thermal units
  - Minimum and maximum power output
  - ✓ Minimum up time
  - Minimum down time
  - Consistency between binary variables

• 
$$\gamma_{k,t}^c - \gamma_{k,t-1}^c = \alpha_{k,t}^c - \beta_{k,t}^c$$
  $2 \le t \le 24$ 

$$\gamma_{k,t}^c - \gamma_{k_0}^c = \alpha_{k,t}^c - \beta_{k,t}^c \qquad t = 1$$



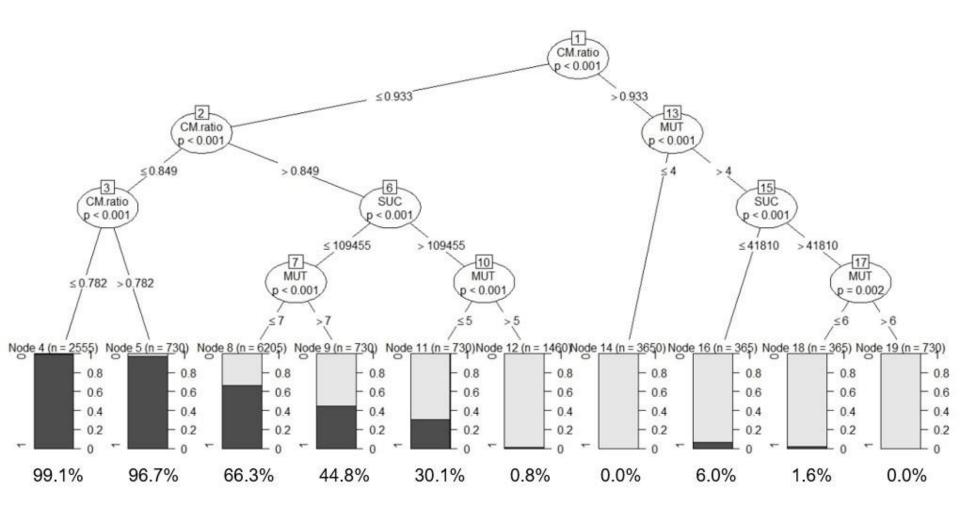
#### **Scenario Data**

- Scenario
  - South-Italy power system
  - ✓ Single year planning horizon
  - ✓ 30% level for RES penetration
- Representative days
  - ✓ Threshold of 1%
  - ✓ 7 representative days
- Initial statuses
  - ✓ Italian power plants
  - Commitment decisions in 2018





#### **Decision Tree**



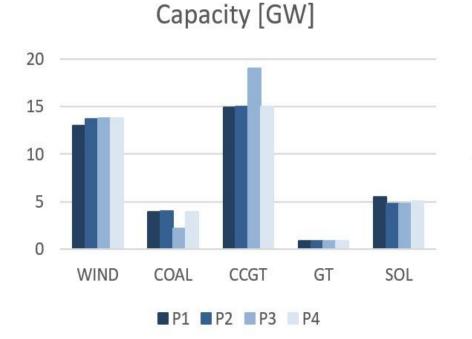


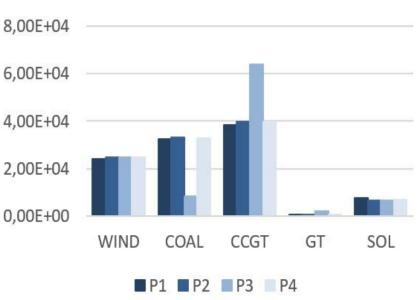
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- Four formulations are compared:
  - ✓ P1 Hourly model (i.e., 8760 values for load, solar and wind power capacity factors are considered)
  - ✓ P2 Representative days are linked, with the initial status of thermal power plants in representative day *c* being equal to the final status in representative day *c* − 1 (i.e.,  $\gamma_{k_0}^c = \gamma_{k,24}^{c-1}$ )
  - ✓ P3 Representative days are not linked, but thermal plants are considered offline at the beginning of each day (i.e.,  $\gamma_{k_0}^c = 0$ )
  - ✓ P4 Representative days are not linked and parameters  $\gamma_{k_0}^c$  are determined with the proposed method









Energy [GWh]





Model	Investment Cost [€]	Production Cost [€]	Start-Up Cost [€]	Total Cost [€]	Total Error	Solution Time [min]
P1	$2.12 \cdot 10^{9}$	$3.04 \cdot 10^{9}$	$4.45 \cdot 10^{7}$	$5.21 \cdot 10^{9}$	—	393.10
P2	2.19 · 10 <sup>9</sup>	$3.11 \cdot 10^{9}$	$3.46 \cdot 10^{7}$	$5.33 \cdot 10^{9}$	2.50%	3.07
P3	$2.21 \cdot 10^{9}$	3.39 · 10 <sup>9</sup>	$44.2 \cdot 10^{7}$	$6.04 \cdot 10^{9}$	16.09%	2.63
P4	$2.16 \cdot 10^{9}$	$3.09 \cdot 10^{9}$	$4.00 \cdot 10^{7}$	$5.29 \cdot 10^{9}$	1.64%	2.57

✓ Formulation *P4* presents both the **highest accuracy** and the **lowest solution time**.

✓ Formulation P4 does not prevent from exploiting the decomposable structure of the expansion planning problem given by the use of disconnected representative days.



## **Future Work**

Integration of the proposed procedure in a detailed model for GTEP:

- Thermal, hydro, wind and solar generation
- Transmission network
- Energy storage systems
- Demand side management devices
- Gas network
- Power-to-gas facilities
- Uncertainty inclusion





## **Thanks for your attention**





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