SMART GRIDS: FROM TRADITIONAL TO MODERNIZED RESILIENT SYSTEMS

Víctor Escala García

Josep Fanals Batllori

Pol Heredia Julbe

Roger Izquierdo Toro

Palina Nicolas

SMART GRIDS

Master's degree in Energy Engineering

Master's degree in Electric Power Systems and Drives



Abstract

This report presents the methodology applied to improve the performance of a transmission network affected by several difficulties. The main purposes of the study have been the increase in the security of supply and the design of a more resilient electrical system to decrease the dependence on interconnection points to external transport networks. Several contingency analyses have been run to achieve these final goals. The project has been divided into four different phases. In the 1st phase, a study of the current system has been performed, leading to the development of the 2nd phase, which has added several transmission lines to improve the security of supply between generation and demand. The 3rd phase has consisted of the addition of two new renewable energy power plants to the electrical system, while the 4th phase has considered the installation of a storage unit and the restoration of a dismantled gas plant. Finally, a Smart Grid Architecture Model (SGAM) has been elaborated to close the project. By simulating several configurations of the transmission network, this study has reached the optimal structure of the system in economic and performance terms. With the new configuration, a more resilient, secure and environmentally friendly electrical system has been achieved, which completely matches the main objectives of smart grids.

CONTENTS

1	Introduction	2				
2	Phase 1	4				
	2.1 Operating costs	. 7				
	2.2 Problem identification	. 8				
	2.3 Solution suggestion	. 9				
3	Phase 2	10				
	3.1 Addition of lines	. 10				
	3.1.1 Formulation	. 10				
	3.1.2 Results	. 12				
	3.2 Rising the voltage level	. 13				
	3.2.1 Description					
	3.2.2 Results					
4	Phase 3	16				
	4.1 Overview					
	4.2 Renewable integration					
	4.2.1 Solar PV					
	4.2.2 Wind					
	4.3 Base case analysis					
	4.4 Contingency analysis					
5	Phase 4	21				
•	5.1 Storage					
	5.2 Dismantled plant					
	5.3 Base case results					
	5.4 Contingency analysis					
	0.4 Containing analysis	. 20				
6	SGAM	27				
	6.1 Scope and objectives	. 27				
	6.2 Mapping	. 28				
7	Conclusions	34				
8	Biographies 35					
9	Code 3					
Bi	Bibliography 52					

1. INTRODUCTION

In modern life, electricity has become a necessity and everyone, households like industries, has to be provided whenever needed. The challenge is that electricity cannot be stored and demand fluctuates. In order to do so, since the early 20th century, the electricity grid from producer to consumer has rapidly increased, building a great network. Nevertheless, even if now the grid is able to provide anyone with electricity, in today's context it is not enough anymore. The growing demand and the climate change crisis pushes toward a more efficient use of energy and greener production. The actual grid, mainly being built a century ago when renewable energy was not in the topic, is long as fossil fuel power plants need to be relatively far from cities and villages and weak in numerical control. This leads to a lot of losses in the transmission and distribution lines, up to 10% [1]. As a response to those issues, the smart grid concept emerged. Its definition varies but overall its aim is to improve and modernize the actual grid in order to improve its efficiency and reduce greenhouse effect in a cost effective way [2]. For doing so, smart grids can use faster data communication thanks to numerical meters, implementation of proximate electricity sources such as household PV, new electricity market for prosumers, wiser selection of high voltage lines, etc. The smart grid concept is relatively new and because testing it in real life is expensive it is first modeled and optimized before being implemented. Moreover, to optimize the conventional grid, this latest has to be well understood.

Many previous works have been conducted, mainly starting from 2008 [3]. They mostly focus on the engineering, energy and computer science aspects. Reports in the field of environmental science are about 7% of all the publications and those relating to economical aspects, although their number is increasing, they remain very low. Studies can be separated in two categories, those related to specific elements of smart grids and those optimizing the whole. In the first case, every element of smart grids is discussed: the power generation with renewable energy sources, transmission and distribution (bidirectional flow), storage, end user, microgrid, market, meters and communications. New technologies can be implemented in order to control and heal the system in an automated way. For example, Automatic Voltage Regulation (AVR) keeps the voltage within the limits, Energy Management System (EMS) ensures stability of every operating point, Automatic Generation Control (AGC) performs optimal load distribution among the generating units, Advanced Metering Infrastructure (AMI) regroups the meters devices, Meter Data Management (MDM) allows fast bidirectional communication of the measures and helps the decision-making, and others like Distribution Management System (DMS), Geographical Information System (GIS), Outage Management System (OMS), Wide Area Management System (WAMS), and Demand Side Management (DSM) [4].

On the other hand, research has been done in order to optimize all those elements [5]. Most of the time, optimization is simplified by focusing on a single objective at a time. The main objectives are energy performance, economical performance or environmental protection. Nevertheless, multi-objectives optimizations are more likely to be accurate. The two main multi-objective optimization models are the weighting factor method and the pareto model, based on the necessary and sufficient conditions logic. Then the decision variables and the constraints are to be added. The decision variables are values that decision-makers have to decide and to do so several multi-criteria methods have been established. For example, Gu et al. wrote a couple of widely accepted evaluation methods

and indicators for CCHP microgrid planning [5]. Within the constraints, we can cite the power balance, generation capacity, transmission capacity between a microgrid and a large grid, location of the microgrid energy storage system,... The minimum operational cost is also a strong constraint. Different algorithms regrouping the objectives, the decision variables and the constraints have been written. The most frequently used is the genetic algorithm, based on Darwin's selection principle, it selects within all the solutions the best ones each generation but others such as the simulated annealing method, the particle swarm optimization and others are also frequently used.

In this work, a simplified conventional network is considered. This initial network, composed of 4 loads, 1 nuclear plant, a dismantled gas plant and an interconnection with an external electricity grid, does not answer the requirement of a basic grid, i.e. providing electricity at all time. The goal is thus to improve this network. The first stage is to analyze the existing grid and point out the weaknesses in order to suggest solutions. The second stage consists of the improvement of the system thanks to the addition of lines and voltage increase. The third phase improves the network by integrating renewable energy sources. The fourth phase rehabilitates the dismantled gas plant and reduces energy dependence by integrating storage and control. Finally, the Smart Grid Architecture Model (SGAM) for the control is designed. For every improvement, a contingency analysis is made, i.e. the network reaction in case one element fails.

2. PHASE 1

A system such as the one displayed in Figure 1 is analyzed. The network operates at the transmission level and feds dispersed demand points that symbolyze distribution grids. The grid has an interconnection with a transmission grid, and at the same time, some power is provided by the nuclear power plant. Initially, there are no renewable power plants nor storage systems, which compromises the security of the grid.

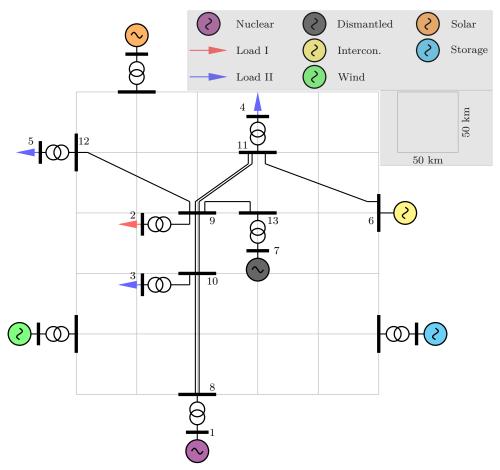


Figure 1. Overview of the network

The first step to analyze the system is to know the demand and the generation profile. In order to model them, the hourly demand and generation data of Spain have been collected [6]. To obtain a typical working day, a statistical analysis has been performed taking into account only the days from the 1st of January to the 31st of March, from Tuesday to Thursday and removing the national holidays. The result then has been normalized. This way, the consumption profile is obtained from the product of the normalized demand and the peak power consumption of 375 MW for load type I and 140 MW for type II. For the generation profile, for simplicity, it has been assumed that the nuclear power plant follows the demand curve, i.e., it is not acting as a constant generator.

Running this primary system, we obtain information of interest. First, Table 1 gathers the per unit voltage for each hour of the normalized day, for each bus while Figure 2 illustrates it. Voltage fluctuations are to be avoided in order to maintain power quality. It is generally accepted that the voltage fluctuates 10% around the nominal per unit voltage [7], i.e. from 0.9 p.u. to 1.1 p.u.. None

of the bus faces an excessive increase in voltage but bus 2, 5 and 12 do drop under the limit. Bus 1 and bus 6 are constant because they directly come from the power plant or the interconnection, they are setting the voltage. It is also visible that the voltage drops more the further the bus is from generation units.

Bus Hour	1	2	3	4	5	6	8	9	10	11	12
0	1.050	0.963	0.989	0.972	0.937	1.000	1.031	0.986	1.000	0.984	0.950
1	1.050	0.974	0.998	0.982	0.952	1.000	1.035	0.995	1.008	0.993	0.963
2	1.050	0.981	1.003	0.988	0.961	1.000	1.037	1.001	1.013	0.998	0.972
3	1.050	0.984	1.005	0.990	0.965	1.000	1.038	1.003	1.015	1.000	0.975
4	1.050	0.985	1.006	0.991	0.966	1.000	1.038	1.004	1.015	1.001	0.976
5	1.050	0.981	1.003	0.988	0.961	1.000	1.037	1.001	1.013	0.998	0.972
6	1.050	0.967	0.992	0.975	0.942	1.000	1.032	0.989	1.003	0.987	0.955
7	1.050	0.938	0.969	0.950	0.905	1.000	1.022	0.965	0.983	0.964	0.920
8	1.050	0.915	0.952	0.931	0.876	1.000	1.014	0.947	0.967	0.946	0.893
9	1.050	0.904	0.944	0.921	0.862	1.000	1.010	0.938	0.960	0.938	0.880
10	1.050	0.900	0.941	0.918	0.856	1.000	1.009	0.935	0.957	0.935	0.875
11	1.050	0.901	0.941	0.919	0.857	1.000	1.009	0.936	0.958	0.935	0.876
12	1.050	0.904	0.944	0.921	0.861	1.000	1.010	0.938	0.960	0.938	0.879
13	1.050	0.906	0.945	0.923	0.864	1.000	1.011	0.939	0.961	0.939	0.882
14	1.050	0.916	0.953	0.931	0.877	1.000	1.014	0.948	0.968	0.947	0.894
15	1.050	0.922	0.957	0.936	0.884	1.000	1.016	0.953	0.972	0.952	0.901
16	1.050	0.925	0.960	0.939	0.889	1.000	1.018	0.955	0.974	0.954	0.905
17	1.050	0.926	0.961	0.940	0.890	1.000	1.018	0.956	0.975	0.955	0.906
18	1.050	0.921	0.957	0.936	0.884	1.000	1.016	0.952	0.972	0.951	0.900
19	1.050	0.903	0.943	0.920	0.860	1.000	1.010	0.937	0.959	0.937	0.878
20	1.050	0.891	0.933	0.910	0.844	1.000	1.005	0.927	0.950	0.927	0.863
21	1.050	0.897	0.939	0.915	0.853	1.000	1.008	0.933	0.955	0.932	0.871
22	1.050	0.921	0.957	0.936	0.884	1.000	1.016	0.952	0.972	0.951	0.900
23	1.050	0.946	0.975	0.957	0.915	1.000	1.025	0.972	0.988	0.970	0.929

Table 1. Voltage profile, in pu, for 24 hours

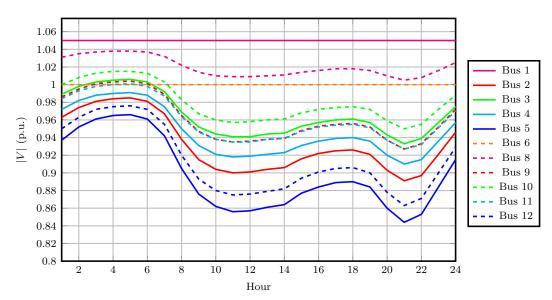


Figure 2. Voltage profile during 24 hours for the initial grid. The low-voltage buses are plotted in solid lines; the high-voltage ones are in dashed lines.

Secondly, the loading of the different lines is gathered in Table 2 and illustrated in Figure 3. The data show the percentage of the used capacity of the line. It is considered overloaded when the load exceeds 80% of the line capacity. Here, only the line 6-11, coming from the interconnection, is overloaded. The 3 lines least loaded are the double lines. Finally, it is interesting to notice that Figures 2 and 3 work as mirrors. When the load decreases, the voltage increases and vice versa.

Load Hour	8-10	10-9	9-11	9-12	11-6
0	26.423	19.999	9.388	30.774	63.562
1	24.385	18.464	8.850	28.312	59.190
2	23.089	17.491	8.517	26.761	56.451
3	22.549	17.087	8.381	26.117	55.319
4	22.458	17.019	8.358	26.009	55.130
5	23.082	17.486	8.515	26.752	56.435
6	25.729	19.475	9.203	29.932	62.065
7	30.615	23.176	10.549	35.928	72.752
8	34.100	25.846	11.559	40.331	80.765
9	35.690	27.073	12.031	42.383	84.443
10	36.296	27.542	12.213	43.173	85.847
11	36.166	27.441	12.174	43.002	85.544
12	35.740	27.111	12.046	42.448	84.558
13	35.480	26.910	11.969	42.110	83.956
14	33.997	25.766	11.528	40.198	80.526
15	33.090	25.069	11.262	39.042	78.435
16	32.569	24.669	11.110	38.381	77.234
17	32.392	24.534	11.059	38.158	76.828
18	33.187	25.143	11.290	39.165	78.657
19	35.865	27.208	12.084	42.610	84.847
20	37.641	28.588	12.620	44.945	88.969
21	36.689	27.847	12.332	43.688	86.759
22	33.143	25.110	11.278	39.109	78.557
23	29.319	22.190	10.183	34.319	69.877

Table 2. Percentual loading of the lines for a full day operation



Figure 3. Representation of the percentual loading of the lines during 24 hours

2.1. Operating costs

Regarding the operating costs, some estimations are made in order to assess the influence of importing energy and the impact of faults on lines and transformers.

First, the cost of importing energy depends on the time zone: valley, flat or peak. The analysis that follows considers a working day, which is precisely the date for which the voltages and loading profiles have been shown in Figures 2 and 3 respectively. The cost of importing the energy is mathematically expressed as:

$$C_{imp} = \sum_{k=1}^{n=24} P_{s,k} c(k), \tag{1}$$

where C_{imp} stands for the cost of importing energy for a full day, k denotes the index of a given hour, n the total number of hours in a day, $P_{s,k}$ the energy provided by the slack bus (interconnection point) in MWh at hour k, and c(k) the cost at a certain hour in ϵ /MWh. This last term is equal to 45ϵ /MWh from 0 to 8 hours, 65ϵ /MWh from 8 to 10, 14 to 18 and 22 to 24 hours, and 90ϵ /MWh from 10 to 14 and 18 to 22 hours.

Equation 1 can be treated as a weighting sum. With the generation data obtained from the timeseries power flow, the total importing cost of importing energy becomes 418753.03 €/day, or about 152.84 M€ for a full year. It is important to note that the study related to the cost of importing energy is decoupled from the fault analysis. This is not a hundred percent realistic, because it could be that a switch trips and hence a line or a transformer is disconnected. Then, it could happen that the interconnection has to provide more power. However, since the probabilities are extremely low, they are discarded when computing this cost.

On the other hand, there are the costs due to faults in transformers or lines. About 0.05 failures per km and year are expected in lines, while transformers are meant to fail 0.15 times a year. The penalty for not providing energy is $180 \in /MWh$. Given that the length of the lines has an impact on its probability of failure, Table 3 shows the length and the subsequent failures per year.

Line	Length (km)	Failures/year
8-10	100.00	5.00
10-9	50.00	2.50
9-11	70.71	3.54
9-12	111.80	5.59
11-6	111.80	5.59

Table 3. Length and failures per year of all active lines

Figure 3 shows that the line connected to the interconnection point operates at a high load. It is critical to note that if line 8-10 fails, the slack should provide all power, but this would result in exceeding the thermal capacity of the line. Thus, if line 8-10 fails, no power can reach the loads.

In the case of line failures, there is a total disconnection time of 2.5 hours, while for transformers it is 8 hours. The expected time that an element will be disconnected in a year is found by multiplying the aforementioned disconnection time by the number of failures that take place during a year. Table 4 displays the yearly disconnection time and explains the consequences spotted by running

Element	Disconnection time (h)	Consequences
Line 8-10	12.50	No load served - divergence
Line 9-10	6.25	No load served - divergence
Line 9-11	8.85	Loads at buses 2, 3 and 5 unserved
Line 9-12	13.98	Load at bus 5 unserved
Line 11-6	13.98	No load served
Trafo 1-8	1.20	No load served - divergence
Trafo 2-9	1.20	Load at bus 2 unserved
Trafo 3-10	1.20	Load at bus 3 unserved
Trafo 4-11	1.20	Load at bus 4 unserved
Trafo 5-12	1.20	Load at bus 5 unserved

the power flow. This will allow to estimate the penalties due to disconnection.

Table 4. Disconnection time and consequences of losing each element

Once the unserved loads and the associated disconnection times are known, the next step has to do with applying the penalty as follows:

$$C_{discon} \approx \sum_{i=1}^{10} \overline{P}_{uns,i} t_{discon,i} C_p, \tag{2}$$

where C_{discon} is the total disconnection cost, i represents the index of the line or transformer with a total of 10 elements proped to be disconnected (see Table 4), $\overline{P}_{uns,i}$ is the mean unserved power, $t_{discon,i}$ the disconnection time, and C_p the penalty cost to apply. Equation 2 is an approximation in the sense that the unserved power varies according to the time of the day. To not overcomplicate the problem, it has been decided to pick a representative value such as the average.

The application of Equation 2 yields a total yearly penalty cost of $4.99 \text{ M} \in$. For the most part, it is due to the disconnection of lines. Meshing more the system would decrease this cost, but on the other side, it would increase the investment cost. Hence, there is a trade-off between cost and reliability.

All the calculations related to costs have not set an inferior limit to the voltages. However, some of them are likely to be unacceptable in reality. The project will proceed to discuss solutions to this issue in the following phases.

2.2. Problem identification

The network modeled presents some serious issues. First, in case of fault, the demand cannot be covered. The network is a ramified line but does not have any interconnection within the system. If a fault occurs, the two branches are not connected and have to support the demand of the remaining part on its own. In the case of the nuclear power plant, it cannot produce enough energy to fulfill all the demand and in the case of the interconnection, if it was to cover all, it would be overloaded.

This leads to the second problem the network faces, there is a risk of overloading. This may happen in case of fault or if the the nuclear power plant shuts down because there is no other source of generation. This could lead to burning hence security and material damage issues. A third drawback is the high impact of the interruptions. As many line are single lines and there are

no multiple connections, only ramifications, a fault has a high chance to directly disconnect the network. Finally, the voltage cannot be kept constant enough. It is usually accepted to fluctuate 10% around the nominal 1 p.u. while in the current transmission network, the voltage reaches almost 0.8 p.u..

2.3. Solution suggestion

As some of the lines present some overloading and demand coverage problems, we suggest improving lines to better ones which are able to transport more power. In order to do so, the critical lines could be changed from single to double lines and/or even change the conductors to thicker cables which allow a larger amount of power flow. Another approach would be to add more lines to the grid to overcome the demand coverage but we must also be aware that when adding new lines to the system we are also increasing the possibility of line failures which may affect interruptibility and cause economic losses to the system.

On the other hand, adding generation points to the system would also help overcome the stated problems in the previous point. If power is more accessible in different locations, the demand can be fulfilled from various points without saturating the most critical lines while evenly distributing the generation. Finally, another solution could be to change the 220 kV existing lines to 400 kV ones in order to allow these to transport higher amounts of power. With this change, only the amount of power transported would be around three times higher than the current one. However, it has to be taken into account that there would have to be an additional transformer to adapt the 220 kV from the interconnection to 400 kV, and the rest of the transformers would have to be replaced to match nominal voltages.

3. PHASE 2

This chapter considers upgrading the system by installing additional lines. In the previous analysis, it was visible that the disconnection of specific lines resulted in an unsolvable problem. Often, no solution was obtained as a result of an exceedingly high demand compared to the available generation. The criteria N-1 was therefore not met, yet it is usually treated as a requirement by control center operators [8], [9].

The central aim of this chapter is to reach a robust grid, where a single failure does not compromise the full system. Several solutions were proposed, such as meshing the network, rising the voltage level, etc. This chapter is devoted to exploring the implications of such proposals. First, a more theoretical explanation describing the formulation is presented, and then, the analysis of results takes place. These results include electrical magnitudes as well as economic data. An optimal configuration is finally provided.

3.1. Addition of lines

Recall that the original system depicted in Figure 1 was for the most part a radial system, in the sense that there was no redundancy in the connection of lines. The operating costs are non-negligible since a significant amount of energy cannot be delivered during faults. To combat this issue, the installation of various new lines (in red) is examined, as shown in Figure 4.

By installing new lines, an additional installation cost has to be considered, but a reduction on the operating costs is expected. It is worth mentioning that not all proposed new lines ought to be installed. Rather, by adding one or two of them, the system could already be operating under the desired conditions. There is a cost-benefit tradeoff, so only a few specific new lines will be selected.

3.1.1. Formulation

The formulation of the topology deserves special attention. Mathematically, it is defined as:

$$\{\mathcal{N} \in \mathcal{P}([n]) \mid |\mathcal{N}| = k\},\tag{3}$$

where \mathcal{N} denotes a subset, that is, possible topology out of the full set of configurations $\mathcal{P}([n])$, $[n] = \{\mathbf{1}_{\sigma_1}, \mathbf{1}_{\sigma_2}, ..., \mathbf{1}_{\sigma_n}\}$ represents the set of elements that cause variations on the topology, and k symbolizes the total amount of possibilities. In the contingency analysis, the set [n] is a function of the state of the elements (mainly, lines). These can be in service or out of service, so can be regarded as boolean variables. The employed notation, of the form $\mathbf{1}_{\sigma_i}$, is 1 if the indicator variable σ_i holds true; otherwise, it is 0. The indicator variable, in the case under study, is simply the state of each particular line.

Note that the whole set \mathcal{P} can be further divided into the set composed of the original lines \mathcal{A} , and the set of new lines \mathcal{B} . Let j indicate the number of lines in the original set \mathcal{A} , and n-j the amount of lines in the new set \mathcal{B} . In the particularized case shown in Figure 4, j=6 and n-j=8.

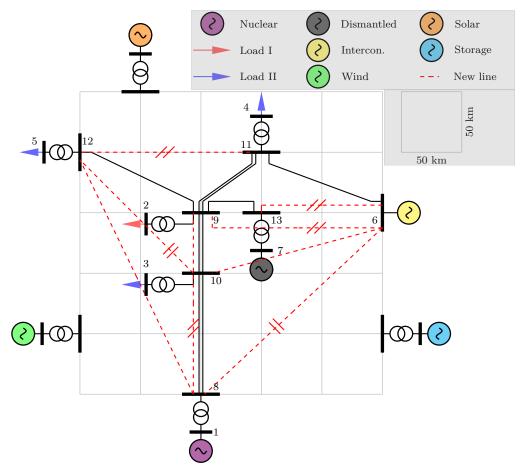


Figure 4. Overview of the network with the addition of lines. Double line if double circuit.

Thus, these subsets are expanded and explicitly become:

$$\begin{cases} \mathcal{A} = \{\mathbf{1}_{\sigma_1}, \mathbf{1}_{\sigma_2}, ..., \mathbf{1}_{\sigma_j}\}, \\ \mathcal{B} = \{\mathbf{1}_{\sigma_{j+1}}, \mathbf{1}_{\sigma_{j+2}}, ..., \mathbf{1}_{\sigma_n}\}. \end{cases}$$
(4)

To perform the contingency analysis, we consider that one of the original lines suffers a fault. For instance, if line i ends up disconnected, $\sigma_i \leftarrow \texttt{false}$ while the rest of original lines $\sigma_r \leftarrow \texttt{true}$ for $r = \{1, 2, ..., i-1, i+1, ..., j\}$.

On the other hand, in the subset \mathcal{B} it is not straightforward to estimate which lines should be connected or disconnected. Perhaps installing a single additional line is enough to meet the N-1 criteria, or maybe more lines are required. Hence, it has been decided that all permutations have to be analyzed. Since σ are boolean variables, the total number k of contingencies to simulate are:

$$k = j2^{(n-j)},\tag{5}$$

so if j = 6 and n - j = 8, this results in k = 1536 different topologies. Thus, a total of 1536 timeseries power flows are computed.

Security analysis in power systems has been traditionally based on the usage of the DC power flow [10], [11]. It offers the advantage of being non-iterative (and consequently, fast), and allows to determine the influence of faults with the usage of transmission factors. However, its solution is

just an approximation. Since the amount of situations remains reasonable, it has been decided to solve them with the typical Newton-Raphson with the default precision of $1 \cdot 10^{-8}$ MVA [12].

The general procedure to perform the contingency analysis is shown in Algorithm 1.

```
Algorithm 1: Pseudocode to solve the contingencies
```

```
Input: net initialized class, j, n
      Output: stored results
 1 Generate permutations \forall \boldsymbol{\sigma}_g where g = [1, 2, ..., 2^{(n-j)}]
 2 for i = [1, 2, ..., j] do
            \sigma_i \leftarrow \mathtt{false}
            \sigma_r \leftarrow \mathsf{true}, where r \neq i and r \leq j
            \mathcal{A} \leftarrow \{\mathbf{1}_{\sigma_1}, \mathbf{1}_{\sigma_2}, ..., \mathbf{1}_{\sigma_j}\}
            for g = [1, 2, ..., 2^{(n-j)}] do
 6
                    [\sigma_{j+1}, \sigma_{j+2}, ..., \sigma_n] \leftarrow \boldsymbol{\sigma}_g
 7
                   \mathcal{B} \leftarrow \{\mathbf{1}_{\sigma_{j+1}}, \mathbf{1}_{\sigma_{j+2}}, ..., \mathbf{1}_{\sigma_n}\}
 8
                   \mathcal{N} \leftarrow \mathcal{A} \cup \mathcal{B}
 9
                   pandapower.timeseries.run_timeseries(\mathcal{N},net)
10
                   Store results
```

Out of all results extracted, an analysis procedure is required in order to determine which configuration meets the N-1 criteria. The requirements are listed below:

- The losses in the lines have to stay below 2%.
- The voltages have to be between a range of $\pm 10\%$.
- All lines have to remain below 80% of its maximum capacity.

Apart from these requirements, an additional criteria has to do with the costs. These will be critical when choosing the best configuration out of all possible ones. Table 5 displays the costs for the lines, depending on their representative voltage level and the number of circuits.

Type	Mean (€)	Min-max range (€)	Median (€)
380-400 kV, 2 circuits	1060919	579771 - 1401585	1023703
380-400 kV, 1 circuits	598231	302664 - 766802	597841
220-225 kV, 2 circuits	407521	354696 - 461664	437263
220-225 kV, 1 circuits	288289	157926 - 298247	218738

Table 5. Network infrastructure costs for the lines. Data from [13].

It has been decided to calculate the costs with mean value.

3.1.2. Results

Out of all results extracted, an analysis procedure is required in order to determine which configuration meets the N-1 criteria. The prechosen configurations are the ones that ensure that in spite of the disconnection of a given element, the power flow remains feasible and the design criterias are met. A total of 39 have satisfied all requirements. Table 6 shows the top 10 cheapest topologies, according to the costs in Table 5.

Identifier	New lines	Infraestructure cost (M€)
19	[6-13, 6-10, 8-9, 10-12]	359.50
214	[6-13, 6-10, 11-12, 8-9]	362.99
77	[6-13, 8-9, 10-12, 9-6]	375.04
49	[6-13, 11-12, 8-9, 9-6]	378.54
80	[6-13, 6-10, 11-12, 8-9, 10-12]	420.63
189	[6-13, 6-10, 8-9, 10-12, 9-6]	420.63
45	[6-13, 6-10, 8-9, 10-12, 8-12]	423.96
70	[6-13, 8-9, 10-12, 9-6]	424.12
157	[6-13, 6-10, 11-12, 8-9, 8-12]	427.46
250	[6-13, 11-12, 8-9, 10-12, 9-6]	436.17

Table 6. Best configurations with the additional lines

The results indicate that the differences in the infrastructure costs are rather minimal. However, installing only four additional lines tends to be the best option. Line 6-13 has been found necessary to ensure a proper operating state of the system under all near-optimal configurations. This makes sense considering that the interconnection usually provides most of the power. Hence, a strong connection to the network is convenient. In fact, the best topology includes two lines connecting the slack bus to the rest of the system are present.

By ensuring that the N-1 criteria is met, we guarantee that all power will be provided independently of which element fails. In reality, this matter may be a bit more complex as two lines could fail simultaneously. Nevertheless, the associated cross-probability is extremely low, and hence, neglected. This implies that no operating costs have to be calculated; it is known beforehand that they will be null as all power is provided for the optimal configuration.

3.2. Rising the voltage level

3.2.1. Description

While installing more lines is a technically feasible strategy, it may not be the most optimal in terms of costs since the voltage level of the grid could be increased. The price to pay for such a change is the replacement of the initial transformers of 25/220 kV and 36/220 kV, for transformers that raise the voltage up to 400 kV (thus, 25/380 kV and 36/380 kV).

In practical terms, raising the voltage may not be as simple as replacing four transformers. For instance, the insulators and the switchgear would have to be changed as well. In the worst scenario, the lines in the towers would have to be more separated, which could cause a replacement of the whole set of supports. To keep the problem arguably simple, it is taken into account that the cost of the transformer station is proportional to the voltage — as can be deduced in [13].

The benefits of increasing the votlage become obvious when observing that the power in a threephase system such as this one is given by:

$$S = \sqrt{3}V_{ll}I,\tag{6}$$

where V_{ll} identifies the line voltage. With this simple relationship, if it is assumed that S remains constant, and V_{ll} increases by a factor of $380/220 \approx \sqrt{3}$, the current would decrease by approx-

imately a factor of $\sqrt{3}$. This means that the active power losses, which are proportional to the square of the current, are roughly reduced by a factor of 3.

Regarding the costs, representative values in ϵ /km are gathered in Table 7. While there is a certain dependence on the type, a generic substation could be taken as reference. Thus, a cost of 42627 ϵ /kV is chosen as an indicator. It is considered that all elements in the substation would have to be replaced to adapt the new voltage level of 380 kV. If there are four substations to be replaced, the total cost for doing so becomes 64.79 M ϵ .

Type	Mean (€/kV)	Min-max range (€/kV)	Median (€/kV)
All substations	42627	24994 - 55508	36755
All GIS substations	46237	29837 - 56017	37449
AIS with 9+ bays	44008	28838 - 56157	41080
AIS with 5-8 bays	35593	19936 - 37251	29021
AIS with 1-4 bays	33192	20276 - 48319	26628

Table 7. Network infrastructure costs for the substation. Data from [13].

3.2.2. Results

The same amount of contingencies as before are computed. Out of the 1536, this time more permutations have become feasible (52 in total). Of course, raising the voltage has a positive effect in this regard, as previously hypothesized. In any case there is a trade-off situation: while the kilometers of new lines may decrease, there is an additional cost due to the new substations. It has to be assessed if the saving due to raising the voltage outweights the cost of installing new substations. Table 8 shows such results for the 10 cheapest options, where the cost of the substations are independent of the topology.

Identifier	New lines	Lines (M€)	Transformers (M€)	Total (M€)
163	[6-13, 8-9, 10-12]	313.92	64.79	378.71
133	[6-13, 11-12, 8-9]	317.41	64.79	382.20
235	[6-13, 8-9, 8-12]	320.75	64.79	385.54
30	[6-13, 8-12, 6-8]	346.07	64.79	410.86
19	[6-13, 6-10, 8-9, 10-12]	359.50	64.79	424.29
214	[6-13, 6-10, 11-12, 8-9]	362.99	64.79	427.78
58	[6-13, 6-10, 8-9, 8-12]	366.33	64.79	431.12
77	[6-13, 8-9, 10-12, 9-6]	375.04	64.79	439.83
169	[6-13, 8-9, 10-12, 8-12]	378.38	64.79	443.17
49	[6-13, 11-12, 8-9, 9-6]	378.54	64.79	443.33

Table 8. Economic results of replacing the substations

A similar conlcusion as before is reached. A new line that connects the interconnection with the dismantled plant is required to meet the planning requirements and achieve an optimal topology. Notice that thanks to the increase in the voltage level, only three new lines should be installed in the optimal configurations. This allows to reduce the cost of these new lines up to 313.92 M \in in the best scenario. However, the extra cost due to the replacement of the transformers implies that the total cost of 378.71 M \in surpasses the minimum cost indicated in Table 6, which is equal to 359.50 M \in .

Therefore, even if both options do not differ much in costs, it would still be better to operate around the nominal voltage of 220 kV. This does not mean, however, that installing more lines is the only feasible solution to confront the technical issues. This may have been true some decades ago, but nowadays, power systems can benefit from the penetration of renewables. Renewable sources of generation can be connected near the loads, which allows to reduce the losses and improve the efficiency of the system. A scenario with renewables is covered in the following chapter.

4. PHASE 3

This chapter covers the inclusion of renewables into the initial grid. It has been verified that the system, as it is configured in the beginning, lacks resiliency. One of its major issues is the presence of only two generation units placed far from the loads. This causes some lines to exceed their loading capacity, while voltages' absolute value may be well below 0.9 p.u.. Adding new lines is a possibility, or rather, a must in order to meet the N-1 criteria. However, it is hypothetized that the performance of the system can still improve with the installation of a wind farm and a PV power plant.

To verify the above-presented hypothesis, the chapter first presents a general overview of the system, along with the sizing of the renewable power plants. Then, the optimal configuration found in the previous chapter is analyzed with the inclusion of renewables to verify the improvement. Finally, a contingency analysis takes place to assess the N-1 criteria.

4.1. Overview

Figure 5 presents the general scheme of the system with the solar and wind plants connected through two parallel lines to nearby buses. Again, new lines that could become suitable for ensuring the robustness of the system are drawn.

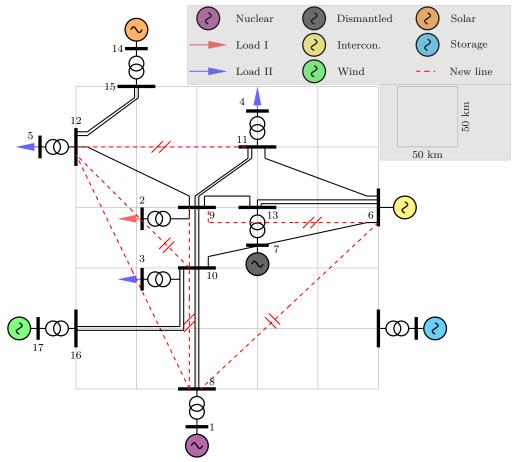


Figure 5. Overview of the network with renewables and the potential addition of lines

The optimal configuration with no renewables, as shown in Table 6, requires the installation of four additional lines: 6-13, 6-10, 8-9, 10-12. The first two are pretty much found in every other topology, as they represent a link with the interconnection. Since the interconnection can theoretically provide any power, it is a good idea to increase the number of associated lines to it. Consequently, lines 6-13 and 6-10 are not considered as potentially attractive new lines, but as already connected lines.

4.2. Renewable integration

In this project, the integration of renewables takes place in the form of solar and wind. The characteristics of each one are particularly detailed. The goal is to characterize their hourly variation of power in relation to the availability of the natural resources. Despite the partial unpredictability that renewables suffer, a representative day in terms of solar irradiance and wind speed has been selected.

4.2.1. Solar PV

First, the PV panels cover an area of 60 ha. This, and some other relevant information is captured in Table 9.

Magnitude	Value	Units
Location	(40.8N; 0.48E)	-
Total area	60	ha
System efficiency	20.7	%
Cell type	Monocrystalline	-
Panel peak power	405	Wp
Panel area	1.95	m^2
Open circuit voltage	37.23	V
Short circuit current	13.87	A

Table 9. Technical characteristics of the PV plant with its panels 405W Deep Blue 3.0 JA Solar Mono [14]

The location has a certain importance to determine the incident irradiation, which is extracted from PVGIS database [15]. In reality, the solar power plant would be located in Tortosa, Catalunya, Spain, according to the given coordinates. The chosen PV panels are of monocrystalline type. Monocrystalline cells generally offer larger efficiencies than polycrystalline cells [16], and although they become more expensive, no cost limitations have been set.

Let n_p denote the total number of panels, G the irradiance, η the efficiency and S_p the area of a single panel. Then, the total generated power follows:

$$P_{PV} = n_p G \eta S_p. \tag{7}$$

It has to be noted that the irradiance G is varying along the day, and it generally has a peak around noon. The data have been selected for the 15th of February. There are a total of $n_p=63692$ panels as it has been assumed that the 60 ha surface is the useful one. Figure 6 shows the hourly irrandiance and power of the PV plant. Clearly, they are proportional, just like it has been already shown in Equation 7.

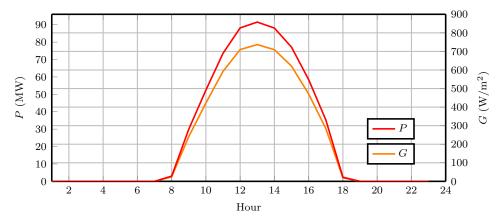


Figure 6. Irradiance and power from the PV plant along a representative day. Data from [15].

The peak in power takes place at hour 13, more or less as expected, and the curves follow a profile that resembles a Gaussian bell. The maximum power generated by the PV plant is 91.5 MW. Although this is a respectable amount, it is roughly one tenth of the load peak. As a consequence, results are not expected to change signficantly. Most likely we will observe slight improvements.

4.2.2. Wind

On the other hand, wind power is proposed as the second renewable alternative. The chosen wind turbines are the Siemens-Gamesa G132-5.0MW. They are rather voluminous turbines, with a rotor diameter of 132 m. Table 10 shows their most notorious technical characteristics.

Magnitude	Value	Units
Location	(42.28N; 3.16E)	-
Total area	80	$_{\mathrm{ha}}$
Nominal power	5.0	MW
Cut-in wind speed	1.5	m/s
Rated wind speed	13.0	m/s
Cut-out wind speed	27.0	m/s
Swept area	13685	$ m m^2$

Table 10. Technical characteristics of the wind power plant with its turbines Gamesa G132-5.0MW [17]

The power provided by a wind turbine is given by:

$$P_{wt} = \frac{1}{2}\rho A C_p v^3,\tag{8}$$

where ρ is the air density of approximately 1.225 kg/m³, A stands for the swept area, C_p is the power coefficient limited to 16/27 (known as Betz's limit [18]), and v represents the wind speed. If the power P_{wt} exceeds the nominal power of the turbine, some power will be curtailed so as not to surpass the machine's ratings. In normal operation, the power coefficient is likely to take values around 0.4.

The available area to install the wind turbines is 80 ha. Nonetheless, wind turbines have to be generously separated one from the others. It has been decided that wind turbines should be separated by a minimum of 5 diameters. Similar separations are adopted in the literature [19]. Thus, a total of two wind turbines are installed. The nominal power of the wind farm becomes

10 MW, which is significantly lower than PV. Despite that, this should not come as a surprise since wind power has a relatively low power density [20].

With this, Figure 7 shows the evolution of the wind speed and the total generated power by the wind farm in an orientative day.

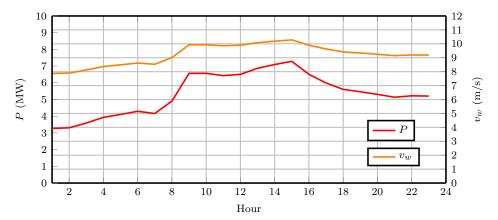


Figure 7. Wind speed and output power from the wind farm. Data from [21].

Notice the cubic relationship between the wind speed and the power. Small increases in wind speed provoke large variations in the output power.

4.3. Base case analysis

This base case analysis refers to the assessment of the influence the installation of renewables has on the results. It is compared to the system analzed before with lines 6-10 and 6-13 permanently added. The goal is not to deduce if the grid meets the N-1 criteria (most likely some other lines would have to be added), but rather compare if the voltages, the loadings of the lines and the losses suffer a certain improvement.

To perform the comparison, the most extreme values have been gathered. They are shown in Table 11 for both cases (with and without renewables).

Attributes	Without renewables	With renewables
V_{min} (p.u.)	0.962	0.968
V_{max} (p.u.)	1.050	1.050
Max. load $(\%)$	41.65	40.76
Max. losses (MW)	14.54	14.43
Correct operation?	Yes	Yes

Table 11. Main results to compare between the grid with and without renewables

The results suggest that generally speaking there is little difference in installing or not renewables. The inclusion of the wind farm and the PV power plant cause the minimum voltage to rise a bit, the maximum loading of the lines to decrease about 1% (although without contingencies it is already low) and the maximum power losses experience a tiny reduction. In the two scenarios, the maximum voltage remains 1.05, which indicates there are no overvoltages. Since all magnitudes are inside the allowed limits, the system is operating correctly in both cases.

Adding renewables to the system comes at a cost. First, there has to be an investment of capital for deploying the wind turbines and the PV panels, apart from power electronics equipment and ancilliary components. Then, a couple of new power lines are added to interconnect them to the system. The slight improvement on the electrical magnitudes will most probably fail at justifying this investment.

However, it is important to note that the major advantage of installing renewables is in its environmental impact. If the power from the interconnection is assumed to come from coal and gas power plants, then installing renewables could become a sensible option. This is especially true when taking into account the actual costs of carbon emissions. Currently at $81.65 \in \text{tonne}$, they have increased by a factor of 3 during the last year [22].

4.4. Contingency analysis

A different story would be the operation with contingencies. Here it is valuable to find out if the presence of distributed renewable generators can help at having to install less extra lines. Hence, the cost could be reduced.

The contingency analysis has been performed similarly to the ones in Tables 6, 8. We take a set of lines that can be potentially installed, and a set of initially powered lines. Contingencies are caused in this latter set to study which new lines have to be connected to meet the requirements. Then, the topologies that ensure the correct operation of the system in spite of which line fails, are selected and their associated costs are calculated. The top 10 most convenient configurations are presented in Table 12.

Identifier	New lines	Infraestructure cost (M€)
24	[8-9, 10-12]	402.71
12	[8-9, 9-6]	406.21
46	[11-12, 8-9]	406.21
8	[8-9, 8-12]	409.54
0	[11-12, 8-9, 10-12]	463.84
4	[8-9, 10-12, 9-6]	463.84
37	[11-12, 10-12, 9-6]	463.84
63	[8-9, 10-12, 8-12]	467.18
20	[11-12, 8-9, 9-6]	467.34
38	[11-12, 8-9, 8-12]	470.67

Table 12. Best configurations with the additional lines

Again, the optimal configuration is the one where apart from lines 6-10 and 6-13, lines 8-9 and 10-12 are added. This coincides with the conclusion we reached in Table 6. Nevertheless, the costs are now higher due to the addition of two new lines to connect the wind and the solar power plants to the system as such. Introducing renewables is not enough to avoid installing the same lines.

In case lines 8-9 and 10-12 are added, the operational costs would be practically zero because we would be meeting the N-1 criteria. To summarize, installing renewables implies an additional cost for the system operator. With time, this could be compensated with the savings in carbon emissions.

5. PHASE 4

While the previous phase included renewables to mitigate the technical problems, another possibility has to do with installing storage systems, rehabilitating a dismantled plant, among others. Certainly, renewables have not been enough to solve the issue of requiring a stronger interconnection. The same number of added lines have been needed in order to meet the N-1 criteria. Hence, it is convenient to evaluate if other elements are capable of improving the energetic independence of the system under study, and at the same time, if it becomes economically appealing.

This chapter is structured in the following manner. First, we detail the characteristics of the storage unit, its operation mode and the expected impact on the system. Then, the dismantled plant is evaluated. Focus is placed on its potential environmental impact rather than on its technical aspects. Once these elements have been defined, results are extracted by running the simulation, including the contingency analysis.

Just as an overview, Figure 8 displays the system with the new elements under consideration.

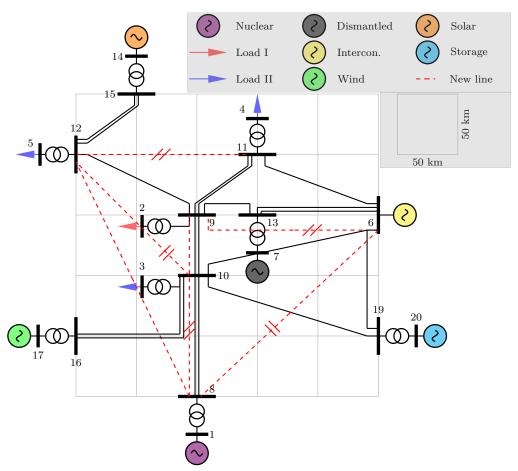


Figure 8. Overview of the network with renewables and the potential addition of lines

5.1. Storage

Storage devices have been conceived as one of the cornerstone class of elements meant to provide flexibility to smart grids [23]. Perhaps in relation to this, electric vehicles (EVs) have been ex-

tensively promoted. Apart from the direct impact on decarbonizing mobility, they can act as a source/sink of energy, which also has an appealing influence on power systems [24]. One of the largest barriers towards adopting batteries has to do with its high investment cost. Although the cost of batteries has been steadily declining for the last two decades [25], there is still a long way to go. Part of the intention regarding this section is to analyze if storage exerts a positive effect on the power flow of the power system under study.

Energy storage options can take many forms. For instance, fast-response devices such as supercapacitors provide peaks of power, yet their accumulated energy is rather low. The same characteristics apply to flywheels, although they store the energy mechanically, not electrically. Another option, likely the most favorable one, are batteries. They store the energy chemically, can provide energy for a sustained period of time, and tend to be easily controlled with power converters. This project conceives the storage unit as a lithium-ion battery, since it offers an attractive trade-off between lifetime, power and energy density, and flexibility of operation, despite its high cost.

Table 13 shows the most notorious characteristics of the chosen lithium-ion battery, along with the corresponding converter.

Magnitude	Units	Value
Capacity	MWh	50
Peak power	MW	10
Round-trip battery η	%	92
Maximum DOD	%	80
Converter η	%	96
Lifetime	cycles	3000
Working temperature	$^{\circ}\mathrm{C}$	-20/55
Specific energy	Wh/kg	133
Orientative LCOE	€/kWh	0.11/0.66

Table 13. Storage system characteristics. DOD: depth of discharge, LCOE: levelized cost of energy. Data from [26]–[28].

Having a lifetime of 3000 cycles means that it is not appropriate to charge and discharge the battery several times per day. If that were to be the case, the energy storage system would last for just a few years at most. On the contrary, if the battery is charged and discharged once per decade, its lifetime could approach a decade. For this reason, the proposed power profile to follow is represented in Figure 9.

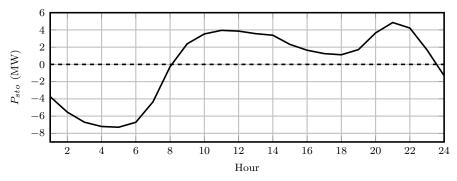


Figure 9. Daily charge and discharge profile for the battery system

This profile resembles the power demand of the loads. This dependence has been established intentionally. In previous analysis, it has been found that during peak hours (around 8 p.m.) the loading of lines is most extreme, and also, the voltages are closer to the lower limit of 0.9 p.u.. Therefore, in general terms, the battery should be discharged during the day and charged at night. According to the chosen sign criteria, positive powers indicate generation, while negative powers denote consumption.

Since the daily demand profile amongst consecutive days do not experience significant differences, it is convenient to always start and finish the day with the same state of charge (SOC). To achieve this, the accumulated area under the curve (in Figure 9, between the solid line and the dashed line) is kept at $E \cdot DOD = 40$ MWh. Besides, the sum of the positive and the negative area yields zero, as desired.

5.2. Dismantled plant

Another possibility to consider in order to reduce the dependence on the interconnection consists of employing the once dismantled power plant placed in bus 7 as indicated in Figure 8. The hypothesis is that this dismantled plant is powered by some kind of fossil fuel (coal, diesel, gas, etc.) or renewable source such as biomass. Thus, even if it could have a positive impact on the electrical magnitudes, its associated environmental impact has to be kept in check.

We start by exploring the diverse energy sources that could power it to extract conclusions regarding costs, emissions, and geopolitical aspects:

- Coal: it is the most polluting of the aforementioned fuels. Even though its emission factor largely depends on the type of coal (lignite, bituminous...), it is in the range of 300 to 360 g CO₂/kWh [29]. Coal power plants are mostly used in countries with availability of natural resources, which is not the case of Spain.
- Diesel: or more generally petroleum, these power plants are more a rarity than an actual
 common choice. They are suitable to be installed in microgrids or islanded systems thank to
 their flexibility and suitability for hybrid systems where they act as backup units [30]. Since
 their chemical proportion between hydrogen and carbon is larger than in the case of coal, its
 emission factor becomes lower; it is around 270 g CO₂/kWh [29].
- Natural gas: it is cleaner than the two previous fuels, with an emission factor of 180 g CO₂/kWh approximately [29]. Power plants powered by natural gas usually take the form of the so-called combined cycle gas turbine (CCGT) power plant. During the previous decades, Spain promoted the installation of combined cycle power plants, which nowadays are unused for the most part [31]. Probably the largest disadvantage of these power plants is the high cost of natural gas, which has fluctuated extensively during the last year [32].
- Biomass: it is considered a renewable source with carbon neutral emissions [29], since the CO₂ emitted during combustion has previously been absorbed from the atmosphere. Then, the emission factor of biomass is 0 gCO₂/kWh. Nowadays, there are already some examples

where the existing facilities have been converted from coal or natural gas to biomass [33].

When it comes to the implementation of the program, it makes no difference to consider one technology or another, as only the output power is relevant. However, this chapter also focuses on the emissions. Therefore, the results section analyzes the results for coal, diesel and natural gas.

The chosen power generation profile for the dismantled plant is depicted in Figure 10. Compared to storage or the demand, here the peaks are more exaggerated. The underlying idea is that the system is more in need of generation during peak hours. And if demand increases, generation should increase by a larger factor to keep the losses and loadings of the lines at satisfactory values. A peak power of 50 MW has been considered, which corresponds to the most critical hour of the day. The plant is supposed to stop operating at night, and its voltage is maintained at 1.02 p.u. since it acts as a PV node.

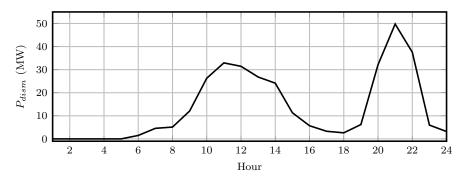


Figure 10. Daily generation profile of the dismantled plant

With storage and the generation of the dismantled plant defined, the simulation can be ran again. The next sections explore the results and draw conclusions from them.

5.3. Base case results

The base case considers the grid as shown in Figure 8, with the black lines representing actual connections. The red lines are only taken into account for the contingency analysis. In any case, it is expected that the presence of the battery and an additional plant should contribute to improving the electrical magnitudes, at least slightly. Table 11 shows this is precisely the case.

Attributes	Phase 2	Phase 3	Phase 4
V_{min} (p.u.)	0.962	0.968	0.971
V_{max} (p.u.)	1.050	1.050	1.050
Max. load $(\%)$	41.65	40.76	39.63
Max. losses (MW)	14.54	14.43	13.89
Correct operation?	Yes	Yes	Yes

Table 14. Main results to compare between the grid without renewables, with renewables, and with storage and the dismantled plant

The results tend to be the ones we could already expect. The minimum voltage rises a bit thanks to the inclusion of more generation sources, which control the voltages at their respective buses and hence ensure the operation around nominal voltages. Similarly, the loadings of the lines are reduced a bit, being the most loaded line at only 39.63%. There could be reasons to think that the system is oversized, because lines do not approach their limits of 80%. While these claims would be full of validity, faults have to be taken into account as well. Redundant connections are a must if the criteria N-1 has to be met.

The maximum losses are again reduced. A clear pattern can be spotted from comparing the results in the three phases. Going from phase 2 to phase 4, results improve each time the grid becomes stronger and more decentralized. These are in essence some of the core characteristics of smart grids. Another central goal of smart grids has to do with decarbonizing the system, which is described below.

Regarding emissions, it is known beforehand that storage, renewables and nuclear do not emit CO₂ gases or equivalent during their operation. The only generation sources to account for emissions are the interconnection and the dismantled plant. In this base case, the dismantled plant produced 332 MWh, whereas the interconnection provided 4686 MWh. The interconnection can be assumed to have an emission factor of 190 kg CO₂-eq/MWh as in the case of Spain [34], [35].

With this, the total emissions for a full-day operation are gathered in Table 15 depending on the technology of the dismantled power plant.

Fuel	Dismantled	Interconnection	Total
	$(tCO_2\text{-eq})$	$(tCO_2\text{-eq})$	$(tCO_2\text{-eq})$
Coal	108.56	890.34	998.90
Diesel	89.64	890.34	979.98
Gas	59.76	890.34	950.10
Biomass	0.00	890.34	890.34

Table 15. Total daily emissions depending on the scenario

The conclusion we reach is that even if there is a non-negligible difference in the emissions caused by the dismantled plant, its overall influence is not that relevant. The interconnection accounts for 90 to 95% of the total CO₂ emissions. The only counteract to that is the integration of nuclear and renewable power plants, which produce about 10.06 GWh daily. This is about twice the energy imported from the slack bus, so in proportion, about 2/3 of the total energy come from clean sources. This is equivalent to saying that the electricity emissions factor of the whole system is roughly 63 kg CO₂-eq/MWh. This is a more than acceptable value that most European countries do not reach (see [36]).

Finally, an economic comparison can be performed according to the related costs of each energy source. These costs include the net present value of the unit-cost of electricity over the lifetime of a system, known as Levelized Cost of Electricity (LCOE), and the CO₂ emission costs. Considering the LCOE of each energy source [37]–[40] and the current CO₂ price [41], the biomass power plant is the cheapest option since it has no CO₂ emission costs associated. It is 16.32% cheaper than natural gas, followed by coal with 35.97%, and diesel with 65.21%.

5.4. Contingency analysis

Just as in previous phases, a contingency analysis is performed to verify if new lines are required, and if so, which ones should be. The marked red lines shown in Figure 8 are the potentially added lines, while the black ones are the alreay established ones. Table 16 presents the top 10 best configurations under faults.

Identifier	New lines	Infraestructure cost (M€)
33	[8-9]	419.49
24	[8-9, 10-12]	477.12
12	[8-9, 9-6]	480.62
46	[11-12, 8-9]	480.62
8	[8-9, 8-12]	483.96
57	[8-9, 6-8]	505.94
0	[8-9, 10-12]	538.25
4	[8-9, 10-12, 9-6]	538.25
37	[11-12, 10-12, 8-12]	538.25
63	[8-9, 10-12, 8-12]	541.59

Table 16. Best configurations with the additional lines with storage and a dismantled plant

The results indicate that, contrary to previous configurations, installing only the additional line from buses 8 to 9 is enough to meet the N-1 criteria. This is certainly a positive result, because in all other cases, at least two lines had to be added. Notice, however, that most configurations are the same as in Table 12. This makes sense since the topology is basically identical, with just minor variations due to the added storage and the dismantled plant.

The total installation cost increases a bit respect to the case where only renewables were added. Of course, new lines have been included to connect the storage unit to the system. Yet, the conclusion has not changed. Moving towards a smart grid, which includes a larger penetration of renewables, batteries, and other elements, supposes a large investment cost. The question is if this invesment ends up being worth it. Multiple reasons could be added into the equation, but what is for sure is that there is a climate emergency to combat.

6. SGAM

This chapter contains the information regarding the use case regarding contingency analysis. The information is presented visually, grouped in tables and figures for a better comprehension. It has to be mentioned that although the central part of this chapter revolves around the definition of a high level use case, details about the primary use cases are also given. They are partially adapted from the European RESOLVD project [42].

6.1. Scope and objectives

Table 17 presents the scope and the objective of the use case.

Attribute	Description
Scope	The scope of the UC is to evaluate the effects of a fault and calculate any overloads based on a computer application that simulates the power system to be prepared for any possible fault.
Objective	Prediction of the impact of faults.
Related Grid Problems Grid faults:	
	• Short-circuits
	Overloads
	• Undervoltages
Relation to other use cases	• PUC 01: Demand and generation forecasting
	The EF can predict demand and generation curves at specific points of the grid.
	• PUC 02: Grid Operation Scheduling
	The Grid Operation Scheduler is capable of providing an optimal schedule of the grid to avoid possible faults and allow the continuity of supply.
	• PUC 03: Grid observability and monitoring
	It permits to program a series of alerts for a better monitoring of the overall status of the grid.
	• PUC 04: Fault detection and localization
	The Fault Detection Application analyses data, comparing them with the expected operation conditions to detect possible faults.
Viewpoint	Technical
Туре	HLUC

Table 17. Scope and objectives of the contingency analysis use case

Table 18 contains the description of the use case.

Short description

Using simulation data, demand and generation forecasting and monitoring systems, the contingency analysis is able to detect and locate faults. When faults are located, they alert the operator. Then, the power flow is computed to determine the impact on the system.

Complete description

Fault detection and data monitoring constantly analyse the system data to detect and then locate faults in the grid. When faults are detected, the operator is notified, in order to start the mitigation process.

The process consists of forecasting the energy supply and demand conditions of the grid on one hand; on the other evaluating the consistency of the grid after the impact of a fault.

For forecasting energy demand and supply conditions, the energy forecaster performs the forecast using smart meters data stored in the meter data management system and weather forecast data provided by the weather forecaster. This falls somewhat outside the scope of the use case, but the central idea is that these actors provide information related to contingencies which is then further processed.

Table 18. Narratives of the contingency analysis

6.2. Mapping

After having defined the general view of the use case, more details are given following the SGAM model. First, Figure 11 shows the component layer.

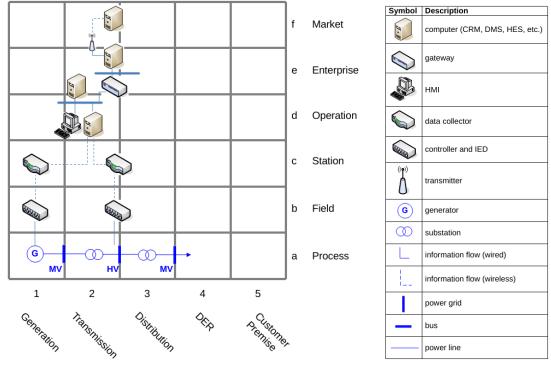


Figure 11. Component layer overview and components legend

Components are for the most part placed between transmission and distribution, as they are the domains where contingencies are assumed to be more common and impactful. The central idea is to gather the data from the field and then be able to process it at the upper zones.

Then, Figures 12 and 13 present the distribution and links between the several actors that intervene.

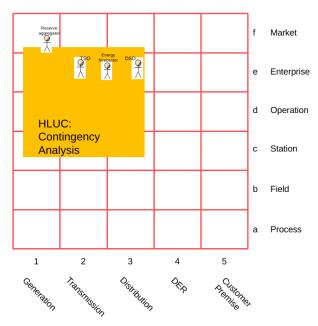


Figure 12. Business layer mapping

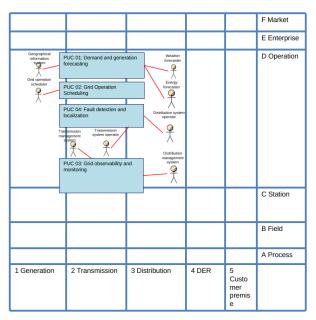


Figure 13. Function layer mapping $\,$

As mentioned before, the actors are generally placed in the generation, transmission and distribution domains. The most relevant zones are the operation, the station and the enterprise. While the operators of the system are the main actors when it comes to business aspects, there are various underlying actors associated to the particular primary use cases (PUC). They several involved

PUCs have been exctracted from [42]. They are defined for the sake of completeness.

PUC 01: Demand and generation forecasting

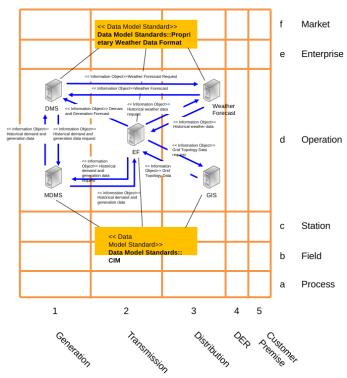


Figure 14. Information layer for PUC 01

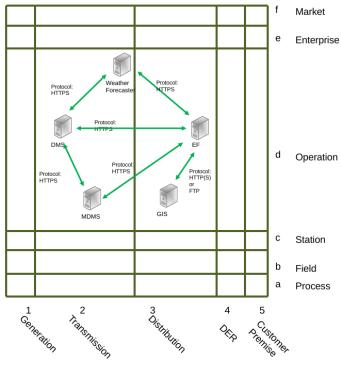


Figure 15. Communication layer for PUC 01

PUC 02: Grid Operation Scheduling

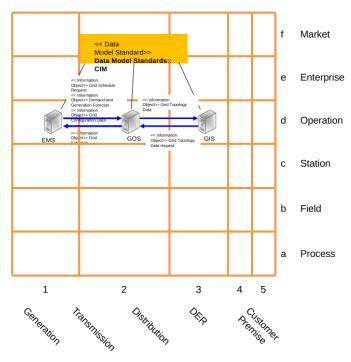


Figure 16. Information layer for PUC 02

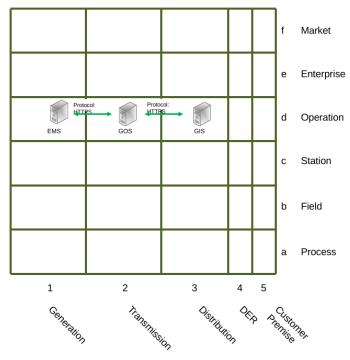


Figure 17. Communication layer for PUC 02 $\,$

PUC 03: Grid ovservability and monitoring

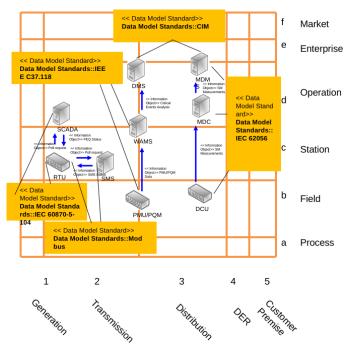


Figure 18. Information layer for PUC 03

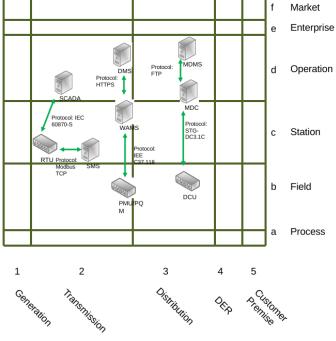


Figure 19. Communication layer for PUC 03

PUC 04: Fault detection and localization

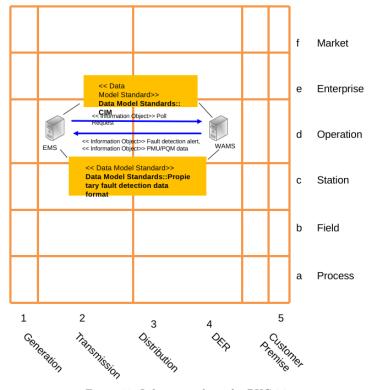


Figure 20. Information layer for PUC 04

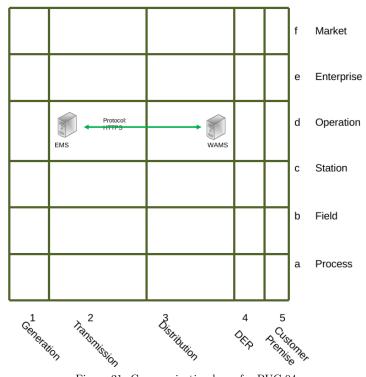


Figure 21. Communication layer for PUC 04

7. CONCLUSIONS

This project has covered the development of a grid, from its traditional structure to a system that integrates renewables and is more in line with the objectives of smart grids. Overall, this work has been splitted in four phases.

The first phase presents the basic system as such. It has been shown that there are two generation sources and four loads. While generation buses are able to set a voltage level, and hence they operate at acceptable levels independently of the hour of the day, loads suffer variations in voltages. This problem is especially acute the further they are from the generation sources. Voltages take values below 0.9 p.u. during peak hours, and although lines are not overloaded, they almost reach the limits.

Phase 2 solves the aforementioned issues by installing more lines. Additional lines increase the reliability of the system — therefore they may allow to operate under the N-1 criteria — but they suppose an increase in cost. Contingency analysis have been performed, where hundreds of combinations have been tested. It was found that up to four extra lines were required to meet the N-1 criteria. Besides, raising the voltage level of the whole grid is another option that has been considered. Nonetheless, it becomes more expensive than installing additional lines.

On the other hand, phase 3 integrates renewables into the system. In more detail, a wind power plant and a photovoltaics power plant are included. They are next to the loads, which helps a bit at raising the voltages, reducing the power losses and the loadings of the lines. However, their power is considerably lower than the nominal power of the nuclear power. Although they do not exert a huge impact, renewables allow to meet the N-1 criteria while having to install less additional lines.

Phase 4 considers the installation of a storage system and a traditional power plant. The capacity of the storage system is just a fraction of the nominal power of the generation sources. Consequently, it had a very small effect on the results. The dismantled power plant becomes a solid selection from the point of view of operation of the system. It is relatively close to the loads, and permits having to add even less power lines to meet all the requirements. Its downside is found in the associated environmental impact. Several cases have been analyzed depending on the fuel it used.

Additionally, the SGAM model has been employed to define a high level use case. In particular, it is focused on contingency analysis. First, a description of the use case, with the objectives and the relation to other use cases has been presented. Then, the component, business and function layers have been filled according to the specifications. Even though primary use cases are not a central part of this work, they have been included to show the full picture of the contingency analysis.

It is also worth mentioning that the simulations have required writting code to compute the power flow of multiple scenarios. The chosen programming language has been Python, due to its flexibility. Mainly, the functions have relied on Pandapower as the main package to call in order to solve the power flow. In this sense, it has been attempted to produce robust code based on object-oriented programming principles. The code is presented in the annex, and it can also be found in GitHub [43].

8. BIOGRAPHIES

Víctor Escala García received the B.S. degree in industrial engineering from ETSEIB (UPC), Barcelona, Spain, in 2020. After that, he is currently finishing the Energy Engineering Master focused on renewable energy. Meanwhile, he has been on an internship with RWE Renewables working on the performance analysis team. He is especially motivated about renewable energy, data science and artificial intelligence.

Josep Fanals i Batllori received the B.S. degree in electrical engineering from Universitat de Girona (UdG), Girona, Spain, in 2020. In 2020, he developed the first open-source version of the holomorphic embedding load-flow method, which is currently available in GridCal. Since 2021 he has been with the Centre d'Innovació Tecnològica en Convertidors Estàtics i Accionaments (CITCEA-UPC), where he has been developing tools for the analysis of power networks. His current interests include numerical methods for the resolution of modern power grids dominated by power electronics.

Roger Izquierdo Toro received the B.S. degree in industrial engineering from ETSEIB (UPC), Barcelona, Spain, in 2020. His bachelor's final project was about the control and simulation of power converters operating in grid-forming mode. Since 2020 he is studying the master's degree in Energy Engineering from UPC. His work experience includes the position of a technical officer from 2019 to 2020 in Citelum Ibérica S.A., which is the company in charge of the public lighting infraestructure in Barcelona. He is currently working as an engineer in Agència de l'Habitatge de Catalunya, which is an institution of the Catalan administration that carries out the restoration and maintenance of social housing.

9. CODE

```
1 import pandapower as pp
2 import pandas as pd
3 import numpy as np
4 \quad {\tt import \ pandapower.control \ as \ control}
5 import pandapower.networks as nw
 6\, import pandapower.timeseries as timeseries
7 import itertools
8 from pandapower.timeseries.data_sources.frame_data import DFData
9 \ \ {\tt from \ pandapower.plotting \ import \ simple\_plot}
10 import time
11
12 import sys
13 import codecs
14
15 \  \, {\tt from \; line\_param\_calc \; import \; calc\_line}
16
18 pd.set_option('display.max_rows', 500)
19 pd.set_option('display.max_columns', 500)
20 pd.set_option('display.width', 1000)
22 \ \ def \ initialize\_net(path\_bus, path\_geodata, path\_line, path\_demand, path\_busload, path\_generation
        , path_busgen, path_trafo, rene=False, path_solar_prof=None, path_wind_prof=None):
23
24
       initialize the grid from the .csv files
25
26
       :param path_bus: path to the bus .csv file
27
       :param geodata: path to the geodata .csv file
       :param path_line: path to the line .csv file
28
29
       :param path_demand: path to the normalized demand .csv file
30
       :param busload: path to the bus-load look up table .csv file
31
       :param path_generation: path to the normalized generation .csv file
32
       :param busgen: path to the bus-generator look up table .csv file \ensuremath{\mathsf{c}}
33
       :param trafo: path to the trafo .csv file
34
       :param rene: boolean for renewables
35
       :return: the net class
36
37
       def create_bus(path_bus, path_geodata):
39
40
           adapts the data from the bus file (if needed)
41
42
           :param path_bus:
43
           :param path_geodata:
44
            :return: the net with the buses added
45
46
47
           df_bus = pd.read_csv(path_bus)
48
           df_geodata = pd.read_csv(path_geodata)
49
50
           net.bus = df_bus
51
52
           # adapt geodata
53
           for ll in range(len(df_geodata)):
54
                indx_bus = pp.get_element_index(net, "bus", df_geodata['name'][11])
                df_geodata['name'][11] = indx_bus
55
56
57
           net.bus_geodata = df_geodata
58
59
            return net
61
62
       def create_line(path_line):
63
64
            adapts the data from the line file
65
```

```
66
             :param path_line:
 67
             :return: the net with the lines added
 68
 69
            df_line = pd.read_csv(path_line)
 70
 71
            for _, line in df_line.iterrows():
                 from_bus = pp.get_element_index(net, "bus", line.from_bus)
 72
 73
                 to_bus = pp.get_element_index(net, "bus", line.to_bus)
 74
 75
                rr, xx, cc, imax = calc_line(line.a,
 76
                                               line.b,
 77
                                               line.c,
 78
                                               line.d.
 79
                                               line.e,
 80
                                               line.max_i,
 81
                                               int(line.parallel),
 82
                                               line.Rca,
 83
                                               line.Dext.
 84
                                               line.kg)
 85
 86
                pp.create_line_from_parameters(net,
 87
                                                 to_bus,
 89
                                                 length_km=line.length,
 90
                                                 r_ohm_per_km=rr,
 91
                                                 x_ohm_per_km=xx,
 92
                                                 c_nf_per_km=cc,
 93
                                                 max_i_ka=imax,
 94
                                                 name=line.name_l,
 95
                                                 parallel=line.parallel,
 96
                                                 in_service=line.in_service)
 97
 98
             return net
 99
100
101
        def create_load(path_demand, path_busload, path_bus):
102
103
             adapts the load files
104
105
            :param path_demand:
106
            :param path_busload:
107
            :param path_bus:
108
             :return: the net with the loads added
109
110
111
            df_demand = pd.read_csv(path_demand)
112
            df_busload = pd.read_csv(path_busload)
113
            df_bus = pd.read_csv(path_bus)
114
115
             # create basic load dataframe
116
             # find the bus index of each load
117
            load_indx = []
118
             for _, load in df_busload.iterrows():
119
                 bus_load = pp.get_element_index(net, "bus", load.bus)
120
                load_indx.append(bus_load)
121
122
             load_indx = pd.DataFrame(load_indx)
            load_indx = load_indx.rename(columns={0: "bus"})
123
124
125
             # load name and peak power
126
            load_name = df_busload['bus']
127
             load_pmw = df_busload['p_mw']
            load_qmvar = df_busload['q_mvar']
128
129
130
             # merge in a full dataframe
131
             headers = ["name", "bus", "p_mw", "q_mvar"]
132
            df_load = pd.concat([load_name, load_indx, load_pmw, load_qmvar], axis=1)
```

```
133
                       df_load.columns.values[0] = "name"
134
135
                        # create time series from the basic load df
136
                        Nt = len(df_demand)
137
                       N1 = len(df_load)
138
                       pmw_ts = np.zeros((Nt, N1), dtype=float)
139
                        qmvar_ts = np.zeros((Nt, N1), dtype=float)
140
                        for i in range(Nt): # number of time periods
141
                                pmw_ts[i,:] = df_load['p_mw'][:] * df_demand['norm'][i]
142
                                qmvar_ts[i,:] = df_load['q_mvar'][:] * df_demand['norm'][i]
143
144
                        # form loads as a static picture (initial time)
145
                        for ll in range(len(df_busload)):
146
                                pp.create\_load(net, bus=load\_indx['bus'][l1], p\_mw=pmw\_ts[0, l1], q\_mvar=qmvar\_ts[0, l1], p\_mvar=qmvar\_ts[0, l1], p\_mvar=qmv
                    11], name=load_name[11], index=int(11))
147
148
                        \label{eq:dfpload_ts} \texttt{df_pload\_ts} \; = \; \texttt{pd.DataFrame(pmw\_ts, index=list(range(Nt)), columns=net.load.index)}
149
150
                        df_qload_ts = pd.DataFrame(qmvar_ts, index=list(range(Nt)), columns=net.load.index)
151
                        ds_pload_ts = DFData(df_pload_ts)
152
                        ds_qload_ts = DFData(df_qload_ts)
                        const_load = control.ConstControl(net, element='load', element_index=net.load.index,
153
                  variable='p_mw', data_source=ds_pload_ts, profile_name=net.load.index)
154
                        const_load = control.ConstControl(net, element='load', element_index=net.load.index,
                  variable='q_mvar', data_source=ds_qload_ts, profile_name=net.load.index) # add the
                  reactive like this?
155
156
                        return net
157
158
159
                def create_generator(path_generation, path_busgen, path_bus):
160
161
                        adapts the generation files
162
163
                        :param path_generation:
164
                        :param path_busgenerator:
                        :param path_bus:
165
166
                        :return: the net with the generators added
167
168
169
                        df_generation = pd.read_csv(path_generation)
170
                        df_busgen = pd.read_csv(path_busgen)
171
                        df_bus = pd.read_csv(path_bus)
172
173
                        # create basic generator dataframe
174
                        # find the bus index of each gen
175
                        gen_indx = []
176
                        for _, gen in df_busgen.iterrows():
                                bus_gen = pp.get_element_index(net, "bus", gen.bus)
177
178
                                gen_indx.append(bus_gen)
179
180
                        gen_indx = pd.DataFrame(gen_indx)
181
                        gen_indx = gen_indx.rename(columns={0: "bus"})
182
183
                        # load name and peak power
184
                        gen_name = df_busgen['bus']
185
                        gen_pmw = df_busgen['p_mw']
                        gen_vpu = df_busgen['vm_pu']
186
187
188
                        # merge in a full dataframe
                        headers = ["name", "bus", "p_mw", "vm_pu"]
189
190
                        df_gen = pd.concat([gen_name, gen_indx, gen_pmw, gen_vpu], axis=1)
191
                        df_gen.columns.values[0] = "name"
192
193
                        # create time series from the basic load df
194
                        Nt = len(df_generation)
                       Ng = len(df_gen)
195
```

```
196
            pmw_ts = np.zeros((Nt, Ng), dtype=float)
197
            for i in range(Nt): # number of time periods
198
                 pmw_ts[i,:] = df_gen['p_mw'][:] * df_generation['norm'][i]
199
200
             # gen structure for 1 t
201
             for ll in range(len(df_busgen)):
202
                 pp.create\_gen(net, bus=gen\_indx['bus'][11], p\_mw=pmw\_ts[0, 11], vm\_pu=gen\_vpu[11],\\
         name=gen_name[11], index=int(11))
203
204
205
             # timeseries
            df_gen_ts = pd.DataFrame(pmw_ts, index=list(range(Nt)), columns=net.gen.index)
ds_gen_ts = DFData(df_gen_ts)
206
207
208
            const_gen = control.ConstControl(net, element='gen', element_index=net.gen.index,
         variable='p_mw', data_source=ds_gen_ts, profile_name=net.gen.index)
209
210
             return net
211
212
213
        def create_generator_rene(path_generation, path_busgen, path_bus):
214
215
             adapts the generation files
216
217
            :param path_generation:
218
            :param path_busgenerator:
219
             :param path_bus:
220
             :return: the net with the generators added
221
222
223
            df_generation = pd.read_csv(path_generation)
224
            df_busgen = pd.read_csv(path_busgen)
225
            df_bus = pd.read_csv(path_bus)
226
            # df_solar_prof = pd.read_csv(path_solar_profile)
227
            # df_wind_prof = pd.read_csv(path_wind_profile)
228
229
            # create basic generator dataframe
230
             # find the bus index of each gen
231
             gen_indx = []
232
             for _, gen in df_busgen.iterrows():
233
                 bus_gen = pp.get_element_index(net, "bus", gen.bus)
234
                 gen_indx.append(bus_gen)
235
236
             gen_indx = pd.DataFrame(gen_indx)
237
             gen_indx = gen_indx.rename(columns={0: "bus"})
238
239
             # load name and peak power
240
             gen_name = df_busgen['bus']
241
             gen_pmw = df_busgen['p_mw']
242
             gen_vpu = df_busgen['vm_pu']
243
244
            # merge in a full dataframe
245
             headers = ["name", "bus", "p_mw", "vm_pu"]
246
             df_gen = pd.concat([gen_name, gen_indx, gen_pmw, gen_vpu], axis=1)
247
            df_gen.columns.values[0] = "name"
248
249
250
             # create time series from the basic load df
251
            Nt = len(df_generation)
252
            Ng = len(df_gen)
253
254
255
             pmw_ts = np.zeros((Nt, Ng), dtype=float)
256
            for gg in range(Ng):
257
                 for i in range(Nt): # number of time periods
258
                    # pmw_ts[i,gg] = df_gen['p_mw'][gg] * df_generation['norm'][i]
259
                     pmw_ts[i,gg] = df_gen['p_mw'][gg] * df_generation.iloc[i,gg+1]
260
```

```
261
                pp.create_gen(net, bus=gen_indx['bus'][gg], p_mw=pmw_ts[0, gg], vm_pu=gen_vpu[gg],
         {\tt name=gen\_name[gg],\ index=int(gg))} \quad {\tt \#\ take\ t=0}
262
263
                 \# gen structure for 1 t
                 # for ll in range(len(df_busgen)):
264
265
                     # pp.create_gen(net, bus=gen_indx['bus'][11], p_mw=pmw_ts[0, 11], vm_pu=gen_vpu[
         11], name=gen_name[11], index=int(11))
266
267
268
             # timeseries
269
             df_gen_ts = pd.DataFrame(pmw_ts, index=list(range(Nt)), columns=net.gen.index)
270
             ds_gen_ts = DFData(df_gen_ts)
271
             const_gen = control.ConstControl(net, element='gen', element_index=net.gen.index,
         variable='p_mw', data_source=ds_gen_ts, profile_name=net.gen.index)
272
273
            return net
274
275
276
        def create_intercon(path_bus):
277
278
            defines the interconnection (slack bus)
279
280
            :param path_bus:
281
             :return: the net with the interconnection added
282
283
284
            df_bus = pd.read_csv(path_bus)
285
286
            # find the slack index
287
             slack_indx = 0
288
            for ll in range(len(df_bus)):
289
                 # slack_indx = pp.get_element_index(net, "bus", bb.name)
290
                 if df_bus['name'][11] == 'intercon':
291
                     slack_indx = pp.get_element_index(net, "bus", df_bus['name'][11])
292
293
            pp.create_ext_grid(net, slack_indx, vm_pu=1.0, va_degree=0)
294
295
             return net
296
297
298
        def create trafo(path trafo):
299
300
             defines the transformers
301
302
            :param path_trafo:
303
             : {\tt return:} \ \ {\tt the \ net \ with \ the \ transformers \ added}
304
305
306
             df_trafo = pd.read_csv(path_trafo)
307
308
             # for trafo in df trafo:
309
             for _, trafo in df_trafo.iterrows():
310
                 hv_bus = pp.get_element_index(net, "bus", trafo.hv_bus)
                lv_bus = pp.get_element_index(net, "bus", trafo.lv_bus)
311
312
313
                 pp.create_transformer_from_parameters(net,
314
315
                                                         lv bus.
316
                                                         trafo.sn_mva,
317
                                                         trafo.vn_hv_kv,
318
                                                         trafo.vn_lv_kv,
319
                                                         trafo.vkr_percent,
320
                                                         trafo.vk_percent,
321
                                                         trafo.pfe_kw,
322
                                                         trafo.i0_percent,
323
                                                         in_service=trafo.in_service)
324
```

```
325 return net
326
327
328
        # create empty network
329
        net = pp.create_empty_network()
330
331
        # buses
332
        net = create_bus(path_bus, path_geodata)
333
334
        # lines
335
        net = create_line(path_line)
336
337
        # loads
338
       net = create_load(path_demand, path_busload, path_bus)
339
340
        # gens
341
        # if rene is False:
342
           # net = create_generator(path_generation, path_busgen, path_bus)
343
344
        net = create_generator_rene(path_generation, path_busgen, path_bus)
345
346
        # interconnection
347
       net = create_intercon(path_bus)
348
349
        # trafos
350
        net = create_trafo(path_trafo)
351
352
        return net
353
354
355
356 def get_Plosses(net, prnt=False):
357
358
        returns the active power losses
359
360
       :param net: full grid
361
       :param prnt: to print the value
362
        :return: total active power losses
363
364
365
        Pll = sum(net.res_line['pl_mw'])
366
        if prnt is True:
367
           print('The active power losses are: ', Pll, 'MW')
368
369
        return Pll
370
371
372 def get_max_loading(net, prnt=False):
373
374
       returns the maximum loading in percentage
375
376
       :param net: full grid
377
        :param prnt: to print the value
378
        :return: maximum percentual loading of the lines
379
380
381
        L_max = max(net.res_line['loading_percent'])
382
        if prnt is True:
383
           print('The maximum loading is: ', L_max, '%')
384
385
        return L_max
386
387
388 def run_store_timeseries(net, identifier):
389
390
        run the timeseries and store the data
391
```

```
392
        :param net: full grid
393
        :param identifier: the name to append to the .xlsx file
394
        :return: nothing, just store in .xlsx
395
396
397
        ow = timeseries.OutputWriter(net, output_path="./Results/Cases/Case_"+identifier,
        output_file_type=".xlsx")
398
399
        ow.log_variable('res_bus', 'vm_pu')
400
        ow.log_variable('res_line', 'loading_percent')
401
        ow.log_variable('res_line', 'pl_mw')
402
        ow.log_variable('line', 'parallel')
403
        \verb|ow.log_variable('load', 'p_mw')| \verb| # try to read the load to calculate then the efficiency| \\
404
        # timeseries.run timeseries(net)
        timeseries.run_timeseries(net, continue_on_divergence=True)
405
406
407
        \mbox{\tt\#} store other data, like state of the lines...
408
        df_lines_states = pd.DataFrame([net.line['name'], net.line['from_bus'], net.line['to_bus'],
        net.line['in_service']])
409
        df_lls = df_lines_states.T
410
        df_lls.to_excel("./Results/Cases/Case_"+identifier+"/lines_states.xlsx")
411
412
        # store diagnostic
413
        diagn = pp.diagnostic(net, report_style='compact')
414
        # df_diagn = pd.DataFrame(diagn)
415
        df_diagn = pd.DataFrame.from_dict(list(diagn))
416
        df_diagn.to_excel("./Results/Cases/Case_"+identifier+"/diagnostic.xlsx")
417
418
        return ()
419
420
421 def perms(n_extra_lines):
422
423
        generate the permutations
424
425
        :param n_extra_lines: number of added lines to combine
426
        :return: list of permutations
427
428
        # lst = ['True', 'False'] * n_extra_lines
429
        lst = [True, False] * n_extra_lines
430
        perms_all = set(itertools.permutations(lst, int(n_extra_lines)))
431
        perms_all = list(perms_all)
432
433
        return perms_all
434
435
436\, def run_contingencies_ts(path_bus, path_geodata, path_line, path_demand, path_busload,
         path_generation, path_busgen, path_trafo, n_extra_lines=0):
437
438
        run contingencies by disconnecting lines for now
439
440
        :param path: path of the datafiles to create the grid
111
        :param n_extra_lines: number of added lines to combine, the last ones in the .csv
442
        :return: store in xlsx files
443
444
        perms_extra = perms(n_extra_lines)
445
446
        net_ini = initialize_net(path_bus, path_geodata, path_line, path_demand, path_busload,
        path_generation, path_busgen, path_trafo)
447
448
        n_cases = len(perms_extra)
449
        n_lines = len(net_ini.line)
450
451
        # disconnect the initial lines and run the cross cases with connecting the others
452
        for ii in range(n lines - n extra lines):
453
            net_2 = pp.pandapowerNet(net_ini)
454
            net_2.line['in_service'][jj] = False
```

```
455
456
            # evaluate the state by connecting the extra lines
457
            for kk in range(n cases):
458
                # copy the net
459
                net_3 = pp.pandapowerNet(net_2)
460
461
                # change state of the line (True/False)
                net_3.line['in_service'][n_lines - n_extra_lines:] = perms_extra[kk][:]
462
463
464
                # run timeseries and store
                run_store_timeseries(net_3, str(jj) + '_' + str(kk))
465
466
                # run_store_timeseries(net_3, str(jj))
467
                # run_store_timeseries(net_3, str(hash(str(jj) + '_' + str(kk))))
468
469
        # return ()
470
        return n_lines, n_extra_lines, n_cases
471
472
473 def process_contingencies(n_lines, n_extra_lines, n_cases):
474
475
        merge all excels into one but different sheets
476
477
        :param n_lines: number of total lines
478
        :param n_extra_lines: number of added lines
479
        :param n_cases: resulting total number of cases
480
        :return: nothing, just store
481
482
483
        dd0 = pd.DataFrame([])
484
485
        nnx = n_cases * (n_lines - n_extra_lines)
        f0_vpu = dd0.to_excel("./Results/All_vpu_" + str(nnx) + ".xlsx")
486
487
        f0_load = dd0.to_excel("./Results/All_load_" + str(nnx) + ".xlsx")
488
        f0_pl = dd0.to_excel("./Results/All_pl_" + str(nnx) + ".xlsx")
489
        f0_diag = dd0.to_excel("./Results/All_diag_" + str(nnx) + ".xlsx")
        f0_line = dd0.to_excel("./Results/All_line_" + str(nnx) + ".xlsx")
490
491
        f0_parallel = dd0.to_excel("./Results/All_parallel_" + str(nnx) + ".xlsx")
492
        f0_pmw = dd0.to_excel("./Results/All_pmw_" + str(nnx) + ".xlsx")
493
494
495
        w_vpu = pd.ExcelWriter("./Results/All_vpu_" + str(nnx) + ".xlsx")
496
        w_load = pd.ExcelWriter("./Results/All_load_" + str(nnx) + ".xlsx")
497
        w_pl = pd.ExcelWriter("./Results/All_pl_" + str(nnx) + ".xlsx")
        w_diag = pd.ExcelWriter("./Results/All_diag_" + str(nnx) + ".xlsx")
498
        w_line = pd.ExcelWriter("./Results/All_line_" + str(nnx) + ".xlsx")
499
500
        w_parallel = pd.ExcelWriter("./Results/All_parallel_" + str(nnx) + ".xlsx")
501
        w_pmw = pd.ExcelWriter("./Results/All_pmw_" + str(nnx) + ".xlsx")
502
503
        for jj in range(n_lines - n_extra_lines):
504
            for kk in range(n_cases):
                fold_path = "./Results/Cases/Case_" + str(jj) + "_" + str(kk)
505
506
                f1_vpu = pd.read_excel(fold_path + "/res_bus/vm_pu.xlsx")
                f1_load = pd.read_excel(fold_path + "/res_line/loading_percent.xlsx")
507
                f1_pl = pd.read_excel(fold_path + "/res_line/pl_mw.xlsx")
508
509
                f1_diag = pd.read_excel(fold_path + "/diagnostic.xlsx")
510
                f1_line = pd.read_excel(fold_path + "/lines_states.xlsx")
511
                f1_parallel = pd.read_excel(fold_path + "/line/parallel.xlsx")
512
                f1_pmw = pd.read_excel(fold_path + "/load/p_mw.xlsx")
513
514
                f1_vpu.to_excel(w_vpu, sheet_name=str(jj) + '_' + str(kk))
                f1_load.to_excel(w_load, sheet_name=str(jj) + '_' + str(kk))
515
                f1_pl.to_excel(w_pl, sheet_name=str(jj) + '_' + str(kk))
516
                f1_diag.to_excel(w_diag, sheet_name=str(jj) + '_' + str(kk))
517
                f1_line.to_excel(w_line, sheet_name=str(jj) + '_' + str(kk))
518
519
                f1_parallel.to_excel(w_parallel, sheet_name=str(jj) + '_' + str(kk))
520
                f1_pmw.to_excel(w_pmw, sheet_name=str(jj) + '_' + str(kk))
521
```

```
522
523
        w_vpu.save()
524
        w_load.save()
525
        w_pl.save()
526
        w diag.save()
527
        w_line.save()
528
        w_parallel.save()
        w_pmw.save()
529
530
531
        return ()
532
533
534
535\, def find_optimal_config(path_diagN, path_lineN, path_loadN, path_plN, path_vpuN, path_parallelN,
          path_pmwN, n_lines, n_extra_lines, n_cases, exclude_lines):
536
537
        Find the optimal configuration of lines considering convergence, losses, voltages, and costs
538
539
        :param path_diagN: path of the diagnostic N cases
540
        :param path_lineN: path of the line N cases \,
541
        :param path_loadN: path of the loading of the lines N cases
542
        :param path_plN: path of the losses through the lines N cases
543
        :param path_vpuN: path of the voltages N cases
544
        :param path_parallelN: path of the number of parallel lines, \ensuremath{\mathtt{N}} cases
545
        : \verb"param" path_pmwloadN": path of the MW of load in the buses, N cases
546
547
        :return: optimal case
548
549
550
        diag = pd.read_excel(path_diagN, sheet_name=None)
551
        line = pd.read_excel(path_lineN, sheet_name=None)
        load = pd.read_excel(path_loadN, sheet_name=None)
552
553
        pl = pd.read_excel(path_plN, sheet_name=None)
554
        vpu = pd.read_excel(path_vpuN, sheet_name=None)
555
        parallel = pd.read_excel(path_parallelN, sheet_name=None)
        pmwload = pd.read_excel(path_pmwN, sheet_name=None)
556
557
558
559
        on_off_all_lines = []
560
        vec_config = []
561
        for name, sheet in diag.items():
562
             # if len(sheet) == 0 or name[0] not in exclude_lines: # if no diagnostic errors
563
             if True or name[0] not in exclude_lines: # if no diagnostic errors
564
             # if True:
565
                 on_off_lines = line[name]['in_service']
566
                 on_off_all_lines.append(on_off_lines)
567
                 n_parallel = parallel[name]
568
                  \texttt{n\_parallel\_subset} \; = \; \texttt{n\_parallel.loc[0, 0:].T} \quad \text{\# only the first row, all are equal} 
569
570
                 # loadings
571
                 loadings = load[name]
572
                 loadings_subset = loadings.loc[0:, 0:]
573
                 conditional_loading = loadings_subset[loadings_subset[:] > 80].isnull().values.all()
574
                 # if True, good, we are below 80%
575
576
                 # vpu
577
                 vpuss = vpu[name]
578
                 vpuss_subset = vpuss.loc[0:, 0:]
579
                 conditional_vpu1 = vpuss_subset[vpuss_subset[:] > 1.10].isnull().values.all()
580
                 conditional_vpu2 = vpuss_subset[0.01 < vpuss_subset[:]][vpuss_subset[:] < 0.90].</pre>
         isnull().values.all() # to avoid the 0.0
581
                 # if both are True, we are good
582
583
                 # active power losses
                 pls = pl[name]
584
585
                 pls_subset = pls.loc[0:, 0:]
586
                 pmws = pmwload[name]
```

```
587
                 pmw_subset = pmws.loc[0:, 0:]
588
589
                 ok losses = True
590
                 kk = 0
591
                 while ok_losses is True and kk < 24: # calculate efficiency at any time
592
                     P_load_all = sum(pmw_subset.loc[kk,:])
593
                     P_loss_all = sum(pls_subset.loc[kk,:])
594
                     P_gen_all = P_load_all + P_loss_all
595
                     eff = P_load_all / P_gen_all
596
                     if eff < 0.98:
597
                         ok losses = False
598
                     kk += 1
599
                     # print('Losses: Pload, Ploss, Pgen, eff: ')
600
                     # print(P_load_all, P_loss_all, P_gen_all, eff)
601
602
                 # if ok_losses is True, we are good
603
604
                full_condition = conditional_loading and conditional_vpu1 and conditional_vpu2 and
         ok_losses
605
                if full condition is True:
606
                     print(name)
607
                     vec_config.append(name)
608
609
                 # print('Condition: load, vpu1, vpu2, loss: ')
610
                 # print(conditional_loading, conditional_vpu1, conditional_vpu2, ok_losses)
611
612
        df_configs = pd.DataFrame(vec_config)
613
        df_configs.to_excel("./Results/OK_configs.xlsx")
614
615
616
617
618 def select_best(path_configs, path_res_lines, path_ini_lines, n_lines, n_extra_lines, n_cases):
619
620
        Find the single optimal configuration, also considering costs
621
622
        :param path_configs: path to the {\tt OK\_configs.xlsx} file
623
        :param path_res_lines: path to the solution for all lines, to read the sheet and states
624
        :param path_ini_lines: to know the distances
625
        : {\tt return:} \ {\tt data} \ {\tt for} \ {\tt the} \ {\tt optimal} \ {\tt configuration}
626
627
628
        # data costs of the lines
        c_2line = 407521 # euro/km
629
        c_1line = 288289 # euro/km
630
631
632
        # store costs
633
        lengths = pd.read_csv(path_ini_lines)['length']
        circuits = pd.read_csv(path_ini_lines)['parallel']
634
635
        costs = []
636
        for ll in range(len(circuits)):
637
            if circuits.loc[11] == 1:
                 costs.append(c_1line * lengths.loc[11])
638
639
             elif circuits.loc[11] == 2:
640
                 costs.append(c_2line * lengths.loc[11])
641
642
        costs = np.array(costs)
643
644
        # use lines operating state
645
        res_configs = pd.read_excel(path_configs)
646
        res_lines = pd.read_excel(path_res_lines, sheet_name=None)
647
        name_configs_prev = list(res_configs.iloc[:,1])
648
649
        # check if the same _yy is in all lines
650
        configs_all = []
651
        for xx in name_configs_prev:
652
          namex = xx.split('_')[1]
```

```
653
            configs_all.append(namex)
654
655
        nice_configs = set(configs_all)
        n_count_max = 0
656
657
        n_{config} = '00'
658
659
        # look max number of occurrences
660
        for nn in nice_configs:
661
            occurr = configs_all.count(nn)
662
            if occurr > n_count_max:
663
                n_count_max = occurr
664
665
        # store the names of the configurations always valid
666
        valid_configs = []
667
        if n_count_max == n_lines - n_extra_lines:
668
            for nn in nice_configs:
669
                occurr = configs_all.count(nn)
670
                if occurr == n_count_max:
671
                    valid_configs.append(nn)
672
673
674
675
676
        # only check 0_xx to get the data
677
        costs_configs = []
678
        dic_configs = {}
679
        for name, sheet in res_lines.items():
680
            nxx = name.split('_')[1]
681
            if nxx in valid_configs:
682
                on_off_lines = np.array(res_lines[name]['in_service'])
683
                c_total = np.dot(on_off_lines, costs)
                dic_configs[nxx] = c_total
684
685
686
687
        names_final = np.array(list(dic_configs.keys()))
        costs_final = np.array(list(dic_configs.values()))
688
689
        # print(dic_configs.keys())
690
        # print(dic_configs.values())
691
        print(names_final)
692
        print(costs_final)
693
694
        # final_configs = pd.DataFrame([np.array(dic_configs.keys()), np.array(dic_configs.values())
        ], columns=['configs', 'costs'])
695
        final_configs = pd.DataFrame([names_final, costs_final])
696
        # final_configs = pd.DataFrame([np.transpose(names_final), np.transpose(costs_final)])
697
        print(final_configs)
698
699
        final_configs.transpose().to_excel("./Results/OPT_configs.xlsx")
700
701
        return ()
702
703
704
705
706
707
708 if __name__ == "__main__":
709
710
        time_start = time.time()
711
712
        # ------ Inputs -----
713
714
        # load paths
715
        # path_bus = 'Datafiles/phII/bus1.csv'
716
        # path_geodata = 'Datafiles/phII/geodata1.csv'
717
        # path_line = 'Datafiles/phII/line2.csv'
718
        # path_demand = 'Datafiles/phII/demand1.csv'
```

```
719
        # path_busload = 'Datafiles/phII/bus_load1.csv'
720
        # path_generation = 'Datafiles/phII/generation1.csv'
721
        # path_busgen = 'Datafiles/phII/bus_gen1.csv'
722
        # path_trafo = 'Datafiles/phII/trafo1.csv'
723
724
        # rising voltage level
725
        # path_bus = 'Datafiles/phII_380kV/bus1.csv'
726
        # path_geodata = 'Datafiles/phII_380kV/geodata1.csv'
727
        # path_line = 'Datafiles/phII_380kV/line2.csv'
728
        # path_demand = 'Datafiles/phII_380kV/demand1.csv'
729
        # path_busload = 'Datafiles/phII_380kV/bus_load1.csv'
730
        # path_generation = 'Datafiles/phII_380kV/generation1.csv'
731
        # path_busgen = 'Datafiles/phII_380kV/bus_gen1.csv'
        # path_trafo = 'Datafiles/phII_380kV/trafo1.csv'
732
733
734
        # ----- phase 3 -----
735
        # phase with rene
736
        # path_bus = 'Datafiles/phIII/bus1.csv'
737
        # path_geodata = 'Datafiles/phIII/geodata1.csv'
738
        # path_line = 'Datafiles/phIII/line1reduced.csv'
739
        # path_demand = 'Datafiles/phIII/demand1.csv'
        # path_busload = 'Datafiles/phIII/bus_load1.csv'
740
        # path_trafo = 'Datafiles/phIII/trafo1.csv'
741
742
743
        # rene
744
        # path_generation = 'Datafiles/phIII/generation_all.csv'
745
        # path_busgen = 'Datafiles/phIII/bus_gen1.csv'
746
747
        # no rene
748
        # path_generation = 'Datafiles/phIII/generation1.csv'
749
        # path_busgen = 'Datafiles/phIII/bus_gen1_norene.csv'
750
751
        # ----- phase 4 -----
752
        path_bus = 'Datafiles/phIV/bus1.csv'
753
        path_geodata = 'Datafiles/phIV/geodata1.csv'
        path_line = 'Datafiles/phIV/line1reduced.csv'
754
755
        path_demand = 'Datafiles/phIV/demand1.csv'
756
        path_busload = 'Datafiles/phIV/bus_load1.csv'
757
        path_trafo = 'Datafiles/phIV/trafo1.csv'
        path_generation = 'Datafiles/phIV/generation_all.csv'
758
759
        path_busgen = 'Datafiles/phIV/bus_gen1.csv'
760
761
762
763
764
765
766
        # ----- Running -----
767
        # define net
768
769
        # net = initialize_net(path_bus, path_geodata, path_line, path_demand, path_busload,
        path_generation, path_busgen, path_trafo)
770
        net = initialize_net(path_bus, path_geodata, path_line, path_demand, path_busload,
        path_generation, path_busgen, path_trafo, rene=True)
771
772
        # run and store timeseries data, for only 1 case
773
        run_store_timeseries(net, '_00storage')
774
775
        # run contingencies
776
        n_lines, n_extra_lines, n_cases = run_contingencies_ts(path_bus, path_geodata, path_line,
        path_demand, path_busload, path_generation, path_busgen, path_trafo, n_extra_lines=6)
777
        process_contingencies(n_lines, n_extra_lines, n_cases)
778
779
        \# n_lines = 16
780
        # n extra lines = 6
781
        \# n_cases = 64
782
```

```
783
         print(n_lines, n_extra_lines, n_cases)
784
785
         # ----- Processing -----
786
787
         # store
788
         nxx = n_cases * (n_lines - n_extra_lines)
        path_diagN = 'Results/All_diag_' + str(nxx) + '.xlsx'
path_lineN = 'Results/All_line_' + str(nxx) + '.xlsx'
789
790
         path_loadN = 'Results/All_load_' + str(nxx) + '.xlsx'
791
792
        path_plN = 'Results/All_pl_' + str(nxx) + '.xlsx'
793
         path_vpuN = 'Results/All_vpu_' + str(nxx) + '.xlsx'
794
         path_parallelN = 'Results/All_parallel_' + str(nxx) + '.xlsx'
795
         path_pmwN = 'Results/All_pmw_' + str(nxx) + '.xlsx'
796
797
         exclude_lines = [6,7]
798
799
        find_optimal_config(path_diagN, path_lineN, path_loadN, path_plN, path_vpuN, path_parallelN,
           {\tt path\_pmwN} \;,\; {\tt n\_lines} \;,\; {\tt n\_extra\_lines} \;,\; {\tt n\_cases} \;,\; {\tt exclude\_lines})
800
         path_configs = 'Results/OK_configs.xlsx'
801
802
         path_lineN = 'Results/All_line_' + str(nxx) + '.xlsx'
803
         path_line_ini = 'Datafiles/phIV/line1reduced.csv'
804
805
         select_best(path_configs, path_lineN, path_line_ini, n_lines, n_extra_lines, n_cases)
806
807
         end_time = time.time()
808
         print(end_time - time_start, 's')
809
810
811
812
        # ----- Others -----
813
814
        # run diagnostic
815
        # pp.diagnostic(net)
816
        # print(net.res_load)
817
        # print(net.res_line)
818
         # print(net.res_gen)
819
820
        # plot
821
        # pp.plotting.simple_plot(net)
822
         # simple_plot(net)
823
824
         # get_Plosses(net, prnt=True)
825
        # get_max_loading(net, prnt=True)
```

Listing 9.1. Main code in Python with the Pandapower library

```
1 import numpy as np
3\, def calc_line(a, b, c, d, e, immax, npar, Rca, Dext, kgg):
5
       calculate r, x, c, and return also Imax
6
7
      :param a: horizontal distance between A1 and C2
8
      :param b: horizontal distance between B1 and B2
9
       :param c: horizontal distance between C1 and A2
10
       :param d: vertical distance between A1 and B1
11
       :param e: vertical distance between B1 and C1
12
       :param immax: max current in A
13
       :param npar: number of parallel lines (1 or 2)
14
       :param Rca: ac resistance in ohm/km
       :param Dext: external diameter in mm
16
       :param kg: factor of roughly 0.8
17
       :return: r, x, c, imax
18
19
       def single_line(a, b, immax, Rca, Dext, kgg):
```

```
21
22
            calculate the R, X, C parameters, also return {\tt Imax}
23
24
            :param a: horizontal distance between A and {\tt C}
25
           :param b: vertical distance between A and B
26
           :param immax: max current in A
27
           :param Rca: ac resistance in ohm/km
28
            :param Dext: external diameter in mm
29
           :param kg: factor of roughly 0.8
30
31
            :return: R, X, C, Imax, in the units desired by pandapower
32
33
34
           # cardinal: https://www.elandcables.com/media/38193/acsr-astm-b-aluminium-conductor-
        steel-reinforced.pdf
35
           # 54 Al + 7 St, Imax = 888.98 A
36
37
           w = 2 * np.pi * 50 # rad / s
38
           Imax = immax * 1e-3 # kA
39
           \# Stot = 547.3 * 1e-6 \# m2, the total section
40
            \# R_ac_75 = 0.07316 * 1e-3 # ohm / m
41
           \# kg = 0.809 \# from the slides in a 54 + 7
42
43
           R_{ac_{75}} = Rca * 1e_{3} * ohm / m, should we correct by temperatures?
44
           Stot = np.pi * Dext ** 2 / 4 * 1e-6 \# m2, the total section
45
           kg = kgg
46
47
           r = np.sqrt(Stot / np.pi) # considering the total section
48
49
           dab = np.sqrt((a / 2) ** 2 + b ** 2)
           dbc = np.sqrt((a / 2) ** 2 + b ** 2)
50
           dca = a
51
52
53
           GMD = (dab * dbc * dca) ** (1 / 3)
54
           GMR = kg * r
55
           RMG = r
56
57
           L = 4 * np.pi * 1e-7 / (2 * np.pi) * np.log(GMD / GMR) # H / m
58
           C = 2 * np.pi * 1e-9 / (36 * np.pi) / np.log(GMD / RMG) # F / m
59
60
61
           # in the units pandapower wants
62
           R_km = R_ac_75 * 1e3 # ohm / km
           X_km = L * w * 1e3 # ohm / km
63
           C_{km} = C * 1e9 * 1e3 # nF / km
65
66
           return R_km, X_km, C_km, Imax
67
68
69
       def double_line(a, b, c, d, e, immax, Rca, Dext, kgg):
70
71
           calculate the R, X, C parameters, also return Imax
72
73
           :param a: horizontal distance between A1 and C2
74
           :param b: horizontal distance between B1 and B2
75
           :param c: horizontal distance between C1 and A2
76
            :param d: vertical distance between A1 and B1
77
           :param e: vertical distance between B1 and C1
78
           :param immax: max current in A
79
           :param Rca: ac resistance in ohm/km
80
           :param Dext: external diameter in mm
81
            :param kgg: factor of roughly 0.8
           :return: R, X, C, Imax, in the units desired by pandapower
83
84
85
           # cardinal: https://www.elandcables.com/media/38193/acsr-astm-b-aluminium-conductor-
        steel-reinforced.pdf
```

```
86
            # 54 Al + 7 St, Imax = 888.98 A
 87
 88
            w = 2 * np.pi * 50 # rad / s
 89
            Imax = immax * 1e-3 * 2 \# kA, for the full line, x2
 90
            \# Stot = 547.3 * 1e-6 \# m2, the total section
 91
            \# R_ac_75 = 0.07316 * 1e-3 # ohm / m
 92
            \# kg = 0.809 \# from the slides in a 54 + 7
 93
 94
            R_ac_75 = Rca * 1e-3 # ohm / m, should we correct by temperatures?
 95
            Stot = np.pi * Dext ** 2 / 4 * 1e-6 # m2, the total section
 96
            kg = kgg
 97
98
100
            r = np.sqrt(Stot / np.pi) # considering the total section
101
102
            da1b1 = np.sqrt((b / 2 - a / 2) ** 2 + d ** 2)
103
            da1b2 = np.sqrt((a / 2 + b / 2) ** 2 + d ** 2)
104
            da2b1 = np.sqrt((c / 2 + b / 2) ** 2 + e ** 2)
105
            da2b2 = np.sqrt((b / 2 - c / 2) ** 2 + e ** 2)
106
            db1c1 = np.sqrt((b / 2 - c / 2) ** 2 + e ** 2)
107
            db1c2 = np.sqrt((b / 2 + a / 2) ** 2 + d ** 2)
108
109
            db2c1 = np.sqrt((b / 2 + c / 2) ** 2 + e ** 2)
110
            db2c2 = np.sqrt((b / 2 - a / 2) ** 2 + d ** 2)
111
            dc1a1 = np.sqrt((a / 2 - c / 2) ** 2 + (d + e) ** 2)
112
113
            dc1a2 = c
            dc2a1 = a
114
115
            dc2a2 = np.sqrt((a / 2 - c / 2) ** 2 + (d + e) ** 2)
116
117
            dab = (da1b1 * da1b2 * da2b1 * da2b2) ** (1 / 4)
118
            dbc = (db1c1 * db1c2 * db2c1 * db2c2) ** (1 / 4)
119
            dca = (dc1a1 * dc1a2 * dc2a1 * dc2a2) ** (1 / 4)
120
121
            rp = kg * r
122
123
            da1a2 = np.sqrt((a / 2 + c / 2) ** 2 + (d + e) ** 2)
124
            db1b2 = b
            dc1c2 = np.sqrt((c / 2 + a / 2) ** 2 + (d + e) ** 2)
125
126
127
            drap = np.sqrt(rp * da1a2)
128
            drbp = np.sqrt(rp * db1b2)
129
            drcp = np.sqrt(rp * dc1c2)
130
131
            dra = np.sqrt(r * da1a2)
132
            drb = np.sqrt(r * db1b2)
133
            drc = np.sqrt(r * dc1c2)
134
135
            GMD = (dab * dbc * dca) ** (1 / 3)
136
            GMR = (drap * drbp * drcp) ** (1 / 3)
137
            RMG = (dra * drb * drc) ** (1 / 3)
138
139
            L = 4 * np.pi * 1e-7 / (2 * np.pi) * np.log(GMD / GMR) # H / m
140
141
            C = 2 * np.pi * 1e-9 / (36 * np.pi) / np.log(GMD / RMG) # F / m
142
143
            # in the units pandapower wants
144
            R_km = R_ac_75 / 2 * 1e3 # ohm / km, like 2 resistances in parallel
145
            X_km = L * w * 1e3 # ohm / km
            C_{km} = C * 1e9 * 1e3 # nF / km
146
147
148
            return R_km, X_km, C_km, Imax
149
150
        if npar == 1:
151
            rr, xx, cc, imm = single_line(a, b, immax, Rca, Dext, kgg)
        elif npar == 2:
152
```

Listing 9.2. Code for the calculation of lines

BIBLIOGRAPHY

- [1] Council of European Energy Resources, Report on power losses, https://www.ceer.eu/documents/104400/-/-/09ecee88-e877-3305-6767-e75404637087, Accessed: 2021-12-14, 2017.
- [2] M. Ahat, S. B. Amor, M. Bui, A. Bui, G. Guérard, and C. Petermann, "Smart grid and optimization," 2013.
- [3] I. Vakulenko, L. Saher, O. Lyulyov, and T. Pimonenko, "A systematic literature review of smart grids," in E3S Web of Conferences, EDP Sciences, vol. 250, 2021, p. 08 006.
- [4] I. Alotaibi, M. A. Abido, M. Khalid, and A. V. Savkin, "A comprehensive review of recent advances in smart grids: A sustainable future with renewable energy resources," *Energies*, vol. 13, no. 23, p. 6269, 2020.
- [5] K. Gao, T. Wang, C. Han, J. Xie, Y. Ma, and R. Peng, "A review of optimization of microgrid operation," *Energies*, vol. 14, no. 10, p. 2842, 2021.
- [6] Red Eléctrica de España, Sistema de información del operador del sistema (esios). perfiles de demanda y generación. https://www.esios.ree.es/es?locale=en, Accessed: 2021-10-25.
- [7] M. Tsili and S. Papathanassiou, "A review of grid code technical requirements for wind farms," *IET Renewable power generation*, vol. 3, no. 3, pp. 308–332, 2009.
- [8] Q. Wang and J. D. McCalley, "Risk and "n-1" criteria coordination for real-time operations," *IEEE Transactions on Power Systems*, vol. 28, no. 3, pp. 3505–3506, 2013.
- [9] A. Gourtani, H. Xu, D. Pozo, and T.-D. Nguyen, "Robust unit commitment with n-1 security criteria," *Mathematical Methods of Operations Research*, vol. 83, no. 3, pp. 373–408, 2016.
- [10] B. Stott, J. Jardim, and O. Alsaç, "Dc power flow revisited," IEEE Transactions on Power Systems, vol. 24, no. 3, pp. 1290–1300, 2009.
- [11] F. Capitanescu, J. M. Ramos, P. Panciatici, D. Kirschen, A. M. Marcolini, L. Platbrood, and L. Wehenkel, "State-of-the-art, challenges, and future trends in security constrained optimal power flow," *Electric Power Systems Research*, vol. 81, no. 8, pp. 1731–1741, 2011.
- [12] L. Thurner, A. Scheidler, F. Schafer, J. H. Menke, J. Dollichon, F. Meier, S. Meinecke, and M. Braun, "Pandapower an open source python tool for convenient modeling, analysis and optimization of electric power systems," *IEEE Transactions on Power Systems*, 2018, ISSN: 0885-8950. DOI: 10.1109/TPWRS.2018.2829021. [Online]. Available: https://arxiv.org/abs/1709.06743.
- [13] Agency for the Cooperation of Energy Regulators (ACER), *Uic report electricity in-fraestructure*. 2015.
- [14] Autosolar, Pallet paneles 405w deep blue 3.0 ja solar mono, https://autosolar.es/panel-solar-24-voltios/pallet-paneles-405w-deep-blue-30-ja-solar-mono, Accessed: 2021-12-01.
- [15] European Comission, *Photovoltaic geographical information system (pvgis)*, https://ec.europa.eu/jrc/en/pvgis, Accessed: 2021-12-01.
- [16] N. Kumar, P. Yadav, and S. Chandel, "Comparative analysis of four different solar photo-voltaic technologies," in 2015 International Conference on Energy Economics and Environment (ICEEE), IEEE, 2015, pp. 1–6.
- [17] Wind Turbine Models, *Gamesa g132-5.0mw*, https://en.wind-turbine-models.com/turbines/768-gamesa-g132-5.0mw, Accessed: 2021-12-01.
- [18] M. Ragheb, "Wind energy conversion theory, betz equation," Wind Energie, 2014.
- [19] J. Bartl, F. Pierella, and L. Sætrana, "Wake measurements behind an array of two model wind turbines," *Energy Procedia*, vol. 24, pp. 305–312, 2012.

- [20] V. Smil, Power density: a key to understanding energy sources and uses. MIT press, 2015.
- [21] NASA Prediction of Worldwide Energy Resources, *The power project*, https://power.larc.nasa.gov/, Accessed: 2021-12-01.
- [22] Investing, Carbon emissions futures overview, https://www.investing.com/commodities/carbon-emissions, Accessed: 2021-12-07.
- [23] B. P. Roberts and C. Sandberg, "The role of energy storage in development of smart grids," *Proceedings of the IEEE*, vol. 99, no. 6, pp. 1139–1144, 2011.
- [24] V. Monteiro, H. Gonçalves, and J. L. Afonso, "Impact of electric vehicles on power quality in a smart grid context," in 11th International Conference on Electrical Power Quality and Utilisation, IEEE, 2011, pp. 1–6.
- [25] M. S. Ziegler and J. E. Trancik, "Re-examining rates of lithium-ion battery technology improvement and cost decline," *Energy & Environmental Science*, vol. 14, no. 4, pp. 1635–1651, 2021.
- [26] F. Díaz-González, A. Sumper, and O. Gomis-Bellmunt, Energy storage in power systems. John Wiley & Sons, 2016.
- [27] F. Lo Franco, A. Morandi, P. Raboni, and G. Grandi, "Efficiency comparison of dc and ac coupling solutions for large-scale pv+ bess power plants," *Energies*, vol. 14, no. 16, p. 4823, 2021.
- [28] M. M. Rahman, A. O. Oni, E. Gemechu, and A. Kumar, "Assessment of energy storage technologies: A review," *Energy Conversion and Management*, vol. 223, p. 113 295, 2020, ISSN: 0196-8904.
- [29] Catalan Office for Climate Change, Practical guide for calculating greenhouse gas (ghg) emissions, https://canviclimatic.gencat.cat/web/.content/04_ACTUA/Com_calcular_emissions_GEH/guia_de_calcul_demissions_de_co2/190301_Practical-guide-calculating-GHG-emissions_OCCC.pdf, Accessed: 2021-12-10.
- [30] M. S. Ismail, M. Moghavvemi, and T. Mahlia, "Techno-economic analysis of an optimized photovoltaic and diesel generator hybrid power system for remote houses in a tropical climate," *Energy conversion and management*, vol. 69, pp. 163–173, 2013.
- [31] AleaSoft, The combined cycle gas turbines and the wind energy in the spanish electricity mix, https://aleasoft.com/combined-cycle-gas-turbines-wind-energy-spanish-electricity-mix/, Accessed: 2021-12-10.
- [32] Trading Economics, Natural gas usd/mmbtu, https://tradingeconomics.com/commodity/natural-gas, Accessed: 2021-12-10.
- [33] Biomass Magazine, Grontmij converts power plant from natural gas to biomass, http://biomassmagazine.com/articles/12122/grontmij-converts-power-plant-from-natural-gas-to-biomass, Accessed: 2021-12-14.
- [34] Red Eléctrica de España, Emissions from the spanish electricity system have fallen by 30 million tonnes in the last 5 years, https://www.ree.es/en/press-office/news/featured-story/2020/06/emissions-spanish-electricity-system-have-fallen-30-million-tonnes-last-5-years, Accessed: 2021-12-11.
- [35] C. C. Spork, A. Chavez, X. Gabarrell Durany, M. K. Patel, and G. Villalba Méndez, "Increasing precision in greenhouse gas accounting using real-time emission factors: A case study of electricity in spain," *Journal of Industrial Ecology*, vol. 19, no. 3, pp. 380–390, 2015.

- [36] European Environmental Agency, Greenhouse gas emission intensity of electricity generation in europe, https://www.eea.europa.eu/ims/greenhouse-gas-emission-intensity-of-1, Accessed: 2021-12-11.
- [37] IRENA, Renewable power generation costs in 2020, https://www.irena.org/publications/2021/Jun/Renewable-Power-Costs-in-2020, Accessed: 2021-12-14.
- [38] S. Roussanaly, M. Vitvarova, R. Anantharaman, D. Berstad, B. Hagen, J. Jakobsen, V. Novotny, and G. Skaugen, "Techno-economic comparison of three technologies for pre-combustion co 2 capture from a lignite-fired igcc," Frontiers of Chemical Science and Engineering, vol. 14, no. 3, pp. 436–452, 2020.
- [39] Freeing Energy, Energy fact solar, wind, and nat gas are, by far, the cheapest ways to generate electricity, https://www.freeingenergy.com/facts/lcoe-cost-comparison-solar-nuclear-wind-g108/, Accessed: 2021-12-14.
- [40] Lazard, Lazard's levelized cost of energy analysis, https://www.lazard.com/perspective/levelized-cost-of-energy-2017/, Accessed: 2021-12-14.
- [41] Ember, Daily carbon prices, https://ember-climate.org/data/carbon-price-viewer/, Accessed: 2021-12-14.
- [42] J. Melendez Frigola, I. Kokos, H. Tuiskula, S. Marksteiner, A. Sumper, R. Gallart, M. Smolnikar, and F. Torrent Fontbona, "Resolvd-renewable penetration levered by efficient low voltage distribution grids. specifications and use case analysis.," 2019.
- [43] Fanals, J., Smartgrid. github repository, https://github.com/JosepFanals/SmartGrid, Accessed: 2022-01-01.