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Nuclear Engineering – Politecnico di Milano

Pagliuca Simone, et al.

my contact, al.'s contacts

Course: Or other indication
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ABSTRACT

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Key-words: Key, Words, Here

NOMENCLATURE

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CONTENTS

1 INTRODUCTION

1.1 Context

We are in 2050

1.2 Case Study

The study will simulate and SMR installed in a prevalently residential area. The reactor will be coupled with a secondary system and the electricity production should be able to cover the entire demand of the area. The model also takes into account the share of renewable energy sources in the area.

1.2.1 Reactor

Our plant consists in an SMR installed in the Joint Research Center (JRC) in Ispra, Northen Italy. The JRC is a research facility of the European Commission, located on the shores of Lake Maggiore. The facility is home to the ESSOR reactor, currently facing decommissioning. The facility has been chosen since it rapresents an areleady nuclearized site which suggests the possibility of a future nuclear plant. Moreover the installation of the SMR could take place, technically, withing the ESSOR reactor considering that the containment building was oversized due to the particular use case of the ESSOR reactor in order to accomodate not only the reactor iteself but also several laboratories.

1.2.2 Secondary Plant

1.3 Objectives

The objective of this work is to determine the feasibility of this installation both technically and economically. On the techincal side of things we foucs on the capability of the system to cover the energy deamnd and to adapt to transient conditions. The economical side of things will assest the minimum total plant cost that would allow for a positive NPV.

2 ENERGY DEMAND AND SUPPLY

At the basis of our work is the knowledge of a rapresentative, realistic demand profile, that would catch the transient of the energy deamnd throughout the year. Given our focus on the residential sector, the overall energy demand can be seen as the sum of the heating/cooling contribution + all the other electricity consuptions for appliances and lightning. Based on the available data we have chosen to model the system based on the distribution of energy classes. Several semplifications will be made through the following derivation but we point out any possible improvment that could be made but was considered out of the scope of this work.

2.1 Methodology

We utilize energy classes to characterize residential homes based on data from the CENED database. In Italy, an APE (Attestato di Prestazione Energetica) is a document that certifies the energy performance of a building. This certification must be issued by a qualified technician and is mandatory for all buildings that are sold or rented. Each residential unit, corresponding to the residence of a single family, requires its own APE. For instance, a building with three apartments will have three separate APEs, one for each apartment.

By analyzing the APE database, we can establish correlations between energy classes and specific building characteristics, such as the presence of photovoltaic (PV) installations or heat pumps. This approach allows for the easy modification of future scenarios by adjusting the distribution of energy classes across the total number of buildings. Our objective is to determine the most relevant characteristics of each energy class and to derive a representative energy demand profile for each class.

From the CENED database [1] we extracted the information about the relevant municipalities of the region, which we derived based on the map of primary substations from the national grid operator (GSE) [2]. The data was also filtered to include only residential buildings, excluding commercial and industrial ones. This was necessary given the large amount of data from the complete regional database. This way an easier to manage csv file was obtained. Note that an API is available to access the database but it was found to be very slow and inefficient.

2.2 Heating and Cooling Demand

For each energy class we have taken a real example of a residential building from the region under study and computed the hourly thermal demand for heating and cooling thoughout the year. This is obtained by means of a professional software (TERMOLOG) that allows for dynamic simulation of the energy system of the building. This computation is based on the UNI EN ISO 52016 standard. We used one reference building per energy class all with the same utlization profile, which is a simplification to be discussed later. The output of the calculation is an hourly profile of the thermal demand for heating and cooling.

2.2.1 Normalization of the Data

The data of each building had to be normilized to be rapresentative of the "average building". To do so we used the "userful heated area" information from the CENED database. This is the area of the building that is actually used for heating and cooling, excluding areas such as garages, attics, and basements. We normalized the reference profiles over their area and then used the average area of the buildings removing the top 99 buildings since some anomalies have been encountered. The average area ended up being $91.37m^2$.
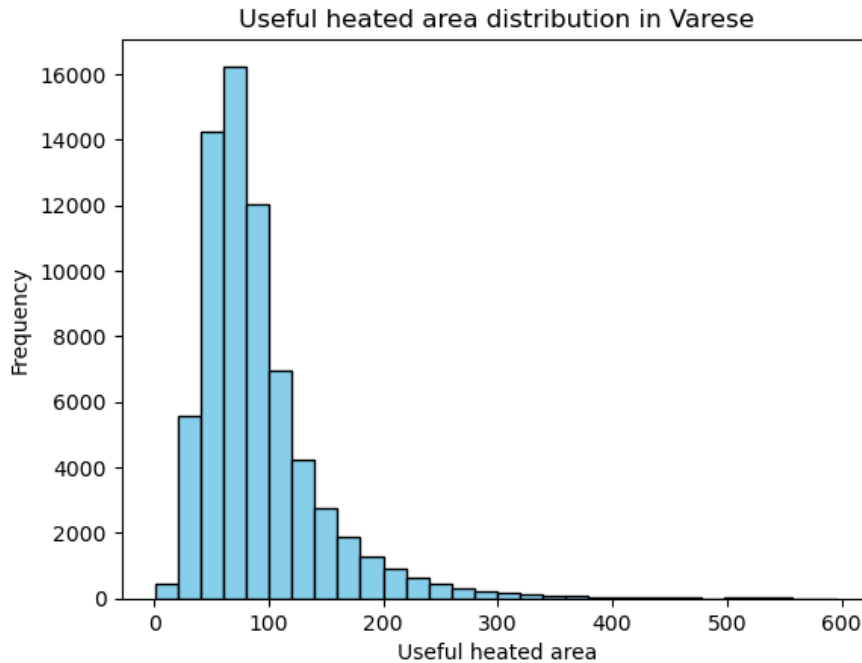
Figure 1: Histogram of the useful heated area of buildings.

2.3 By class distributions

To be able to simply simulate different scenarios we have identified characteristics of each energy class so that we can just modify the energy calss distribution and all the parameters needed to determine the energy demand will follow.

2.3.1 Heat Pump distribution and Cooling Demand

We identified the share of heat pumps per energy class. We also determined the share of buildings that have a cooling system installed, which as of today comes up to 9.58%. This is a very low share and is to be expected as the region under investigation is quite chill during summer.
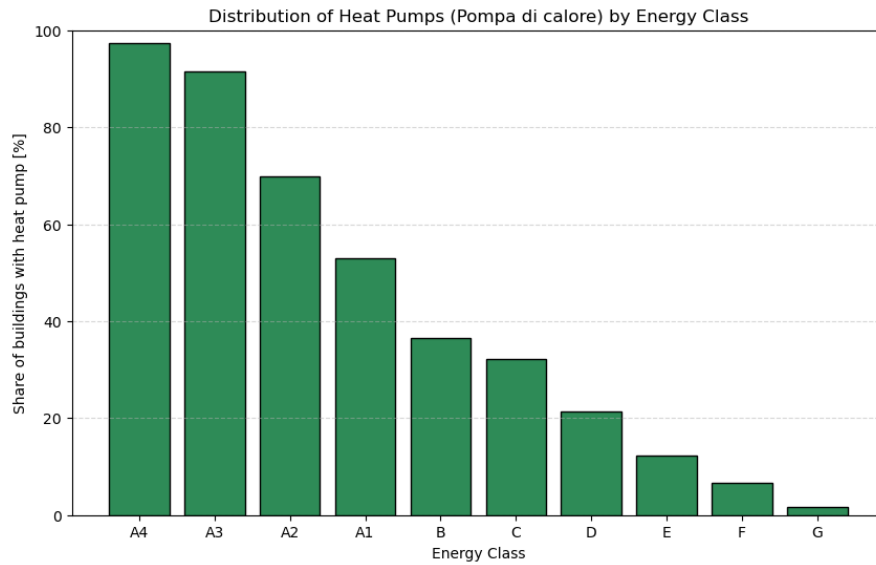
Figure 2: Distribution of heat pumps across energy classes.

```
1   df = pd.read_csv(path, dtype=str) # Read the data
2   energy_classes = ["A4", "A3", "A2", "A1", "B", "C", "D", "E", "F", "G"]
3   impianto_cols = [col for col in df.columns if 'TIPO_IMPIANTO' in col]
4   has_pdc = df[impianto_cols].apply(lambda row: row.str.lower().str.contains('pompa
        di calore', na=False).any(), axis=1)
5   classe = df['CLASSE_ENERGETICA']
6   total_by_class = classe.value_counts()
7   pdc_by_class = classe[has_pdc].value_counts()
8   total_by_class = total_by_class.reindex(energy_classes, fill_value=0)
9   pdc_by_class = pdc_by_class.reindex(energy_classes, fill_value=0) # Reorder by
        class
10  pdc_percentage = (pdc_by_class / total_by_class * 100).fillna(0) # Calculate share
```
Listing 1: Code to extract heat pump data from the CENED database.

```
1   df = pd.read_csv(path, dtype=str) # Read the data
2   sup_raffrescata = df['SUPERF_UTILE_RAFFRESCATA'].astype(float)
3   cooling_count = len(sup_raffrescata[sup_raffrescata > 0]) # Count how many
        buildings have cooling
4   cooling_percentage = (cooling_count / len(df)) * 100 # Percentage of buildings with
         cooling
```
Listing 2: Code to extract cooling system data from the CENED database.

2.3.2 Photovoltaic distribution

We identified the share of buildings with photovoltaic systems installed.

```
1   df = pd.read_csv(path, dtype=str) # Read the data
2   has_pv = df['CONSUMI_SOLARE_FOTOVOLTAICO'].astype(float) > 0
3   classe = df['CLASSE_ENERGETICA']
```

```
4    total_by_class = classe.value_counts() # Total buildings per class
5    pv_by_class = classe[has_pv].value_counts() # Buildings with PV per class
6    pv_percentage = (pv_by_class / total_by_class * 100).fillna(0) # Calculate
         percentage
```

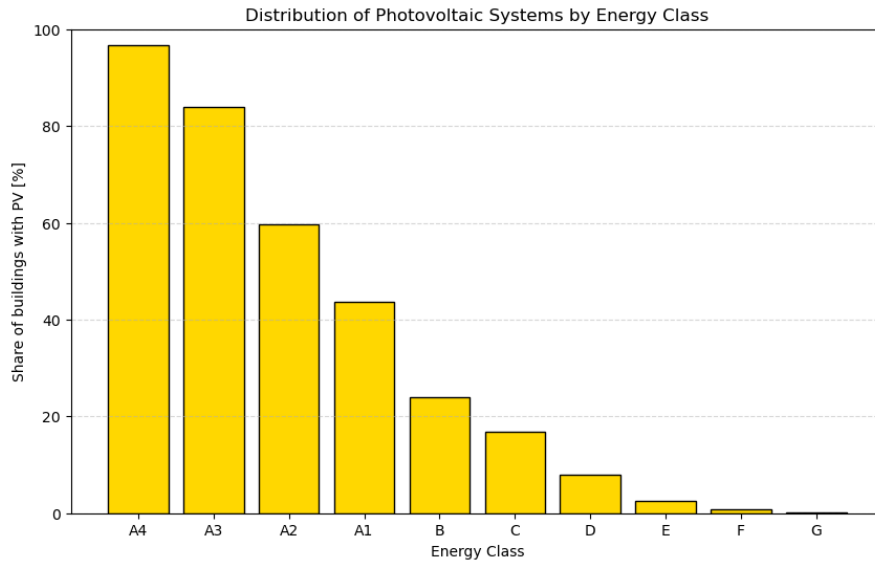Listing 3: Code to extract photovoltaic data from the CENED database.

Figure 3: Distribution of photovoltaic system sizes across energy classes.

Another interesting information we gathered is the relative size of the photovoltaic system. Using data from the CENED database: Exported Electricity, Imported Electricity and In situ consumption.

$$\text{SCF: Self-Consumption Factor} = 1 - \frac{\text{Import}}{\text{Export} + \text{In situ consumption}} \tag{1}$$

$$\text{OSF: Over-Sizing Factor} = \frac{\text{Export}}{\text{Export} + \text{In situ consumption}} - 1 \tag{2}$$

Mind that this data comes from the CENED 2.0 calculation system which uses a non dynamic simulation of the energy system and neglects any storage.

Class	SCF (%)	OSF (%)
A4	61.51	3.96
A3	44.08	34.20
A2	31.68	267.82
A1	25.57	1004.33
B	19.29	947.18
C	15.51	419.08
D	9.16	102.67
E	3.92	116.66
F	1.15	-68.68
G	0.24	-95.72

Table 1: SCF and OSF indicators by Energy Class.

The extreme overdimensioning on some classes can be explained with hands-on experince: To boost the energy class of those in the middle of the scale (betwwen C and E), it is common that contractors choose to an over sized photovoltaic system. This effect was exagerated during the "superbonus" (2021-2023) period which required a mandatory jump of 2 energy classes to be eligible for the tax deduction.

2.4 Convertion from thermal to primary energy demand

The dynamic simulation outputs the thermal demand of the building. To estimate the primary energy needed we used average efficiencies of the heating and cooling systems.

$$\dot{Q}_{\text{fossil}} = \frac{\dot{Q}_{\text{th}}}{\eta_{\text{fossil}}} \tag{3}$$

$$P_e = \frac{P_{\text{th, heating}}}{\text{COP}_h} + \frac{P_{\text{th, cooling}}}{\text{COP}_c} \tag{4}$$

2.5 Other Electricity Demand

Electrical demand for appliances and lighting is modeled equal for all classes. The definition was manual, hour by hour for an average day in the following "seasons":

- Winter Weekday
- Winter Weekend
- Summer Weekday
- Summer Weekend
- Winter Holiday
- Summer Holiday

Moreover we added an absence factor to take into account when people go on holiday massively during august.

The profile is defined by considering the following contributions:

- standby and always on appliances

- lighting

- appliances

During the summer the need for a cooling fan was added. Moreover we added a parameter to rapresent the share of buildings with an induction system installed.

2.6 Total Grid Demand

In the end we choose an energy class distribution to rapresent a future scenario. Given that we know the reference demand-production profile of each building we can sum them up by weighting on their supposed share to obtain the total reference demand which is then multiplied by the number of buildings in the region to obtain the total demand.

2.6.1 Comparison and Tuning

To verify our method is acceptable we compare the total demand by the data available online, in particual we compared it to the average annual deamnd, in the residential sector for the province of Varese [3], which happends to be $1931kWh$ in 2023, with a decreasing trend in the last years ($2016kWh$ in 2022 and $2119kWh$ in 2021).

2.6.2 Possible Improvments

It is defenitly possible and could give better results if the data for each class was obtained by a set of real building from the region instead of just one. Ideally one would use a statistically signfinicant set of buldings and vary the utilization profile as well. Moreover, it would be beneficial to wheight the simulation data on the real energy requirments of those same buildings, which is a much more complex task since it would require extensive data acquisition during the year. While this is feasible for electrical demand (most operators give to the customer a detailed bill with hourly consumption), it is much more difficult to get the same data for fossil fuel consumption. At last, the normalization has been done with the heated area an not with the volume (which is much more indicative of the energy requirment of a building) given the on-field knowledge that the volume is usually approximatly obtained by a the rule of thumb of multiplying the heated area by a factor of 2.7 or 3 in most cases. From the simulation we have neglected humidification and dehumidification electricity needs since these plants are rearly installed today, a better evaluation would evaluate the possible future share of air treatment plants in the residential sector.

Photovoltaic: We have neglected any solar-thermal system installed since their are considered not very impactful and their impact is complex to evaluate.

Convertion thermal to primary: use time dependent efficiency for heat pumps that take into account the ambient temperature, this could be accomplished considering that our code uses hourly data.

Other Electricity Demand: This is without a doubt the most complex contribution to evaluate, expecially considering that it is not easy to imagine how the electricity consumption will evolve in 25 years. It could be improved by collecting a statistically signficant number of examples from the region on the type of appliances installed and their consumption.

Overall: the methodology of using class distribution as a way of evaluating the energy demand is considered a good approach, for the best results we shall gather consumption data for both electric and fossil fuels for a statistically significant number of buildings for each energy class.

3 PLANT MODELLING

4 ECONOMIC STUDY

REFERENCES

[1] Regione Lombardia, *Database CENED+2 - Certificazione ENergetica degli EDifici*, `https://www.dati.lombardia.it/Energia/Database-CENED-2-Certificazione-ENergetica-degli-E/bbky-sde5`, Accessed: 2025-03-25, 2025.

[2] Gestore dei Servizi Energetici (GSE), *Mappa Interattiva delle Cabine Primarie*, `https://www.gse.it/servizi-per-te/autoconsumo/mappa-interattiva-delle-cabine-primarie`, Accessed: 2025-04-14, 2025.

[3] Acquirente Unico S.p.A., *Electricity residential demand*, `https://www.consumienergia.it/portaleConsumi/it/energia-elettrica-utenze-domestiche.page`, Accessed: 2025-04-14, 2025.