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MODULE 1_MAPPING H&C DEMAND AND RES POTENTIAL

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Introduction to Mapping



- Georeferenced data are an important tool for city- and district-level energy planning
- They allow matching energy demand with potential supply from renewable or sustainable sources in a specific area
- They are particularly useful for high-level planning since they allow identifying areas with a good potential for action but additional evaluations are always needed



Mapping in SO WHAT



- The SO WHAT tool includes mapping of geographical distribution of the heating and cooling demand and of the potential energy production from renewable energy sources
- Indeed, mapping these two distributions is important in order to evaluate:
 - the possibility of feeding the recovered heat to residential/tertiary buildings through district heating/cooling networks
 - the possibility to integrate the recovered energy with potential further heat or electricity produced from a renewable source



Mapping H&C Demand



- It is needed to determine the space distribution of heating, cooling and domestic hot water demand over a given area, and its time distribution
- Two approaches are available







Top-Down Approach

- Works at pixel level
- Maps current annual H&C-DHW useful energy demand
- Requires aggregated input data (at city or district scale)
- Disaggregates the total value based on building/population density data
- Monthly/daily distribution of annual demand can be calculated based on Heating and Cooling Degree Days









Bottom-Up Approach

- Works at building level
- Maps hourly and annual profile of H&C-DHW useful energy demand
- Requires input data at building level (cadaster or similar)
- Available in a simplified and in a complete version, depending on available information

Use¤	Construction- Period¤	Number of Buildings¤	GFA• (m²)¤	Heating-Demand- (kWh/year)¤	Cooling-Demand∙ (kWh/year)¤	DHW-Demand- (kWh/year)¤
Residen			279,7			
tial×	Pre-1945×	534×	72×	20,346,771×	473,960×	3,888,831×
Residen			59,4			
tial×	1945-1969×	71×	63×	4,000,259×	86,496×	826,536×
Residen			29,70			
tial×	1970-1979×	32×	3×	1,689,811×	71,039¤	412,872×
Residen			61,73			
tial×	1980-1989×	97×	4×	3,760,774×	218,466≍	858,103×
Residen			87,44			
tial×	1990-1999×	52×	1¤	3,499,476×	259,553×	1,215,430×
Residen			37,48			
tial×	2000-2010×	34×	7¤	951,556×	138,860×	521,069×
Residen			19,67			
tial×	Post-2010×	14×	7¤	500,967×	67,495×	273,510×







Mapping RES Potential



- Determining spatial distribution of renewable source availability
- For sources with significant variability in time (e.g. solar, wind), determining also time distribution









Physical Potential

Geographical Potential

Technical Potential

Local Potential



Mapping Thermal RES Potential



Energy source's	Algorithms	Inputs	External database [unit] to provide	
type	[unit]	[unit]	default values for algorithms' input	
Solar thermal energy E _{pot,solar}	E _{pot,solar} = Ir × Technical suitability × Efficiency × Available Area [MWh/yr]	 Ir =Total solar irradiance under flat plane angle [kWh/m²] Technical suitability [-]: assumed 40% (1); Efficiency [-]: assumed 75% (2); Available area= Footprint of buildings [m²], Other suitable area [m²]. 	 <i>Ir</i> [kWh/m²]: PVGIS database (https://re.jrc.ec.europa.eu/pvg _tools/it/tools.html#MR); Footprint of buildings [m²]: cadaster database or OpenStreetMap (http://www.openstreetmap.or g) as an open data alternative; Other suitable area [m²]: Corine Land cover (https://land.copernicus.eu/pan -european/corine-land- cover/clc2018). 	



Mapping Thermal RES Potential



Energy source		Algorithms	Inputs	External database [unit] to provide default
		[unit]	[unit]	values for algorithms' input
Biomass en from for PE _{FT}	nergy estry	$PE_{FT} = FA \times NAI_{reg}$ $\times PE$ [MWh/yr]	 <i>FA</i> = Forest area of specific forest type: Forest cover [ha], Protected areas or other spatial constrains [ha], Global ecological zones [ha]; <i>NAI_{reg}</i> = Average stem-wood net annual increment per region [tonnes_{DM}/(ha yr)]: GEZ – Temperate forests: 3.0 to 4.0 GEZ – Boreal forests: 1.5 to 2.5 GEZ – Subtropical forests: 0.9 <i>PE</i> = Primary energy production for every type of forest [MWh/ton_{DM}]: 19.0 to 19.2 	 Forest cover [ha]: Corine Land cover (2018) (<u>https://land.copernicus.eu/pan-european/corine-land-cover/clc2018</u>); Protected areas or other spatial constrains [ha]: Natura2000 areas (2016) (<u>http://ftp.eea.europa.eu/www/natura2000</u> <u>/Natura2000_end2016_Shapefile.zip</u>); Global ecological zones [ha]: Global ecological zones (2010) (<u>http://www.fao.org/geonetwork/srv/en/main.home?uuid=2fb209d0-fd34-4e5e-a3d8-a13c241eb61b</u>);

Similar correlations for biomass energy from agriculture



Mapping Thermal RES Potential



Energy source	Algorithms	Inputs	External database [unit] to provide default
	[unit]	[unit]	values for algorithms' input
Deep geothermal energy $H_{tech,heat,x km}$ (150m-7km)	$P_{tech,heat,x \ km}$ $= \frac{T_{x \ km} - T_{r}}{T_{7km} - T_{r}} P_{tech,heat,7km}$ [W] $H_{tech,heat,x \ km} = P_{theory}$ $\times lifetime$ [J]	 T_{x km} = Rocks' temperature at a specific depth [°C]; T_{7 km} = Rocks' temperature at 7 km of depth [°C]; T_r = Temperature of reinjected water [°C]; Lifetime: assumed 8760 hours a year; 	 Spatial constrains: EES-Natura200 areas (http://ftp.eea.europa.eu/www/natur a2000 /Natura2000_end2016_Shapefile.zip) Technical potential deep geothermal energy: database from https://www.thermogis.nl/en/map- viewer; https://map.mbfsz.gov.hu/geo_DH/ T_{x km}, T_{7 km} [°C]: database from https://www.thermogis.nl/en/map- viewer

Similar correlations for shallow geothermal energy



Mapping Electrical RES Potential



Energy source	Algorithms	Inputs	External database [unit] to provide
	[unit]	[unit]	default values for algorithms' input
Photovoltaic energy $E_{pot,ph}$	E _{pot,ph} = Ir × Technical suitability × Panel Efficiency × Electrical Efficiency × Available Area [kWh/yr]	 Ir =Total solar irradiance under flat plane angle [kWh/m²] Technical suitability [-]: assumed equal to 40% (1); Panel Efficiency [-]: assumed equal to 30% Invalid source specified.; Electrical Efficiency [-]: assumed to be equal to 17% Available area= Footprint of buildings [m²], Other suitable area [m²]. 	 Ir [kWh/m²]: PVGIS database (http://re.jrc.ec.europa.eu/PVGI <u>S5-release.html</u>); Footprint of buildings [m²]: cadaster database or OpenStreetMap (http://www.openstreetmap.or g) as an open data alternative; Other suitable area [m²]: Corine Land cover (https://land.copernicus.eu/pan -european/corine-land- cover/clc2018).

Mapping Electrical RES Potential



Energy source	Algorithms	Inputs	External database [unit] to provide
	[unit]	[unit]	default values for algorithms' input
Wind energy at a specific height <i>h</i>	$E_{h} = 8760 \eta_{m} \eta_{el} \eta_{aux}$ $\sum_{w=w_{low}}^{w_{high}} P_{h}(w) f(w)$ $[kWh/yr]$ $f(w)$ $= \left(\frac{\beta}{\eta}\right) \left(\frac{w}{\eta}\right)^{\beta-1} e^{-\left(\frac{w}{\eta}\right)^{\beta}}$ $[-]$	 8760 yearly operative hours of the aerogenerator [h]; <i>ρ</i> = air density [kg/m³]; <i>A</i> = area of the aerogenerator that is orthogonal to wind's direction [m²]; <i>w</i> = wind speed [m/s]; <i>C_p</i> = power coefficient [-]; <i>f</i>(<i>w</i>) = Weibull distribution [-]; <i>β</i> = shape parameter [-]; <i>η</i> = scale parameter [-]. Height at which the wind turbine is placed; <i>η_m</i> mechanical efficiency of the device [-]; <i>η_{el}</i> electrical efficiency of the device [-]; <i>η_{aux}</i> auxiliaries' efficiency [-]. 	 β and η [-]: CENER database (https://globalwindatlas.info/) Spatial constrains: EES- Natura200 areas (https://www.eea.europa.eu/da ta-and-maps/data/natura- 9/natura-2000-spatial- data/natura-2000-shapefile-1).



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THANK YOU FOR YOUR PARTICIPATION

SOWHAT TEAM

