MIT Portugal

Modeling Supply and Demand Dynamics in Energy Systems *A\$图前缀综合*春春春春\$P\$\$R*A\$图前缀综合*春春春春\$P\$ Carlos A. Santos Silva → 🏴 🔄 🖨 🗑 🗦 🤸 → 🎧 🛇 🏶 👁 🚨 🛒 📖 🗐 → 🟴 🐼 🚗 🖾 💂 🐾 ☼▮░░◪ሥ◮⑤◬७፠◬◬◬▧▮R€▧☼▮░░◪ሥ◮⑤◬७፠◬◬ጜ▧▮R

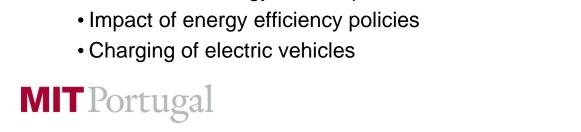
Planning Future Energy Systems

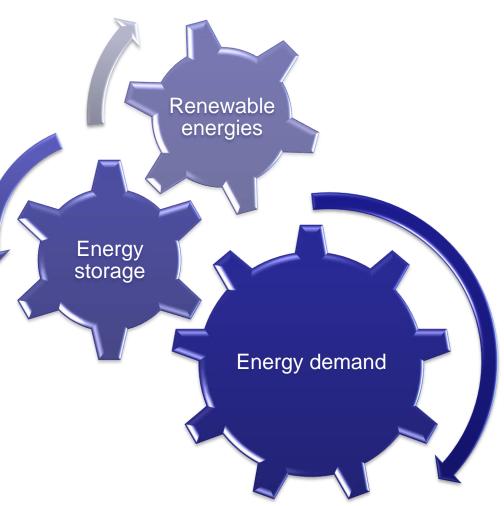
To design sustainable energy, several options must be considered:

- Renewable resources
- Energy storage
- Consumer behavior
- Energy efficiency
- Alternative transportation fuels (biofuels, electricity, others)

To design effectively, the interactions between the possible options must be accounted for:

- Intermittency of renewable resources
- Evolution of energy consumption





Available Tools

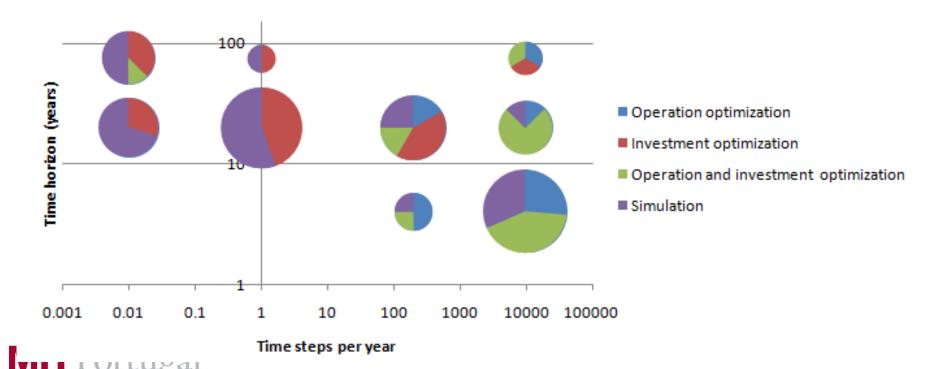
Tool	Туре								
	Simulation	Scenario	Equilibrium	Top-down	Bottom-up	Operation optimisation	Investment optimisation		
AEOLIUS	Yes	-	-	.=:	Yes				
BALMOREL	Yes	Yes	Partial	-	Yes	Yes	Yes		
BCHP Screening Tool	Yes	-		-	Yes	Yes	1 		
COMPOSE	<u>_</u>	-	-	_	Yes	Yes	Yes		
E4cast	-	Yes	Yes	-	Yes	-	Yes		
EMCAS	Yes	Yes	_	-	Yes	-	Yes		
EMINENT		Yes	_	-	Yes	=			
EMPS	2	_	_	_	5.00	Yes	12		
EnergyPLAN	Yes	Yes	<u></u>	12	Yes	Yes	Yes		
energyPRO	Yes	Yes	. — .		7.770	Yes	Yes		
ENPEP-BALANCE	-	Yes	Yes	Yes	_	=	_		
GTMax	Yes	-	-	-	-	Yes	10000 1 — 1		
H2RES	Yes	Yes		_	Yes	Yes	-		
HOMER	Yes	-	_	_	Yes	Yes	Yes		
HYDROGEMS	-	Yes		32	-	-	-		
IKARUS		Yes			Yes		Yes		
INFORSE	_	Yes	_	_	-	Ξ	- -		
Invert	Yes	Yes	_		Yes	_	Yes		
LEAP		Yes		- V	Yes				
	Yes		- Yes	Yes	Yes	- 1- 1- 1- 1- 1- 1- 1- 1- 1- 1- 1- 1- 1-	Yes		
MARKAL/TIMES	-	Yes		Partly		\$ 7 \$			
Mesap PlaNet	-	Yes	ī	-	Yes		-		
MESSAGE	7.	Yes	Partial	7.	Yes	Yes	Yes		
MiniCAM	Yes	Yes	Partial	Yes	Yes	÷	-		
NEMS		Yes	Yes	-	ā		Ξ.		
ORCED	Yes	Yes	Yes	-	Yes	Yes	Yes		
PERSEUS	_	Yes	Yes	-	Yes	-	Yes		
PRIMES	_	-	Yes		_	<u>-</u>	(<u>*</u>		
ProdRisk	Yes	-	2	12	_	Yes	Yes		
RAMSES	Yes	-	7.	1.71	Yes	Yes	-		
RETScreen	-	Yes	7	1.T	Yes	1 .7 1	Yes		
SimREN	-	 -	-	: - :	-	₩.	-		
SIVAEL	-	-	-	-	-	-	-		
STREAM	Yes	-	-	-	-	-	=		
TRNSYS16	Yes	Yes	-	-	Yes	Yes	Yes		
UniSyD3,0	_	Yes	Yes	_	Yes	2	_		
WASP	Yes	_	_	_	_	-	Yes		
WILMAR Planning Tool	Yes	_	<u>- 1</u>	_	_	Yes	<u>-</u>		



Analysis of energy planning tools

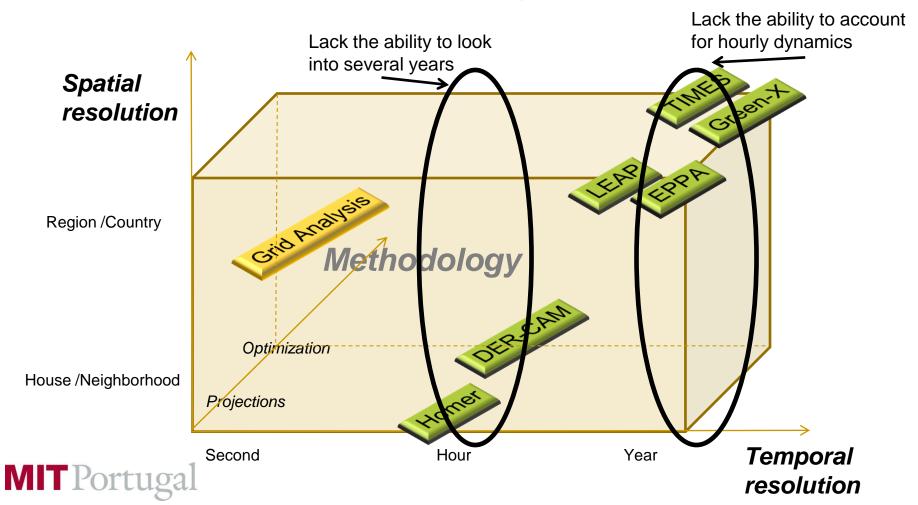
Analysis of 84 energy models showed that:

- Simulation and Invesment optimization models are generally used for simulation of medium and long-term case studies with low resolution
- Operation optimization and Operation and investment optimization models are generally applied to case studies with higher time-resolution than 1h.

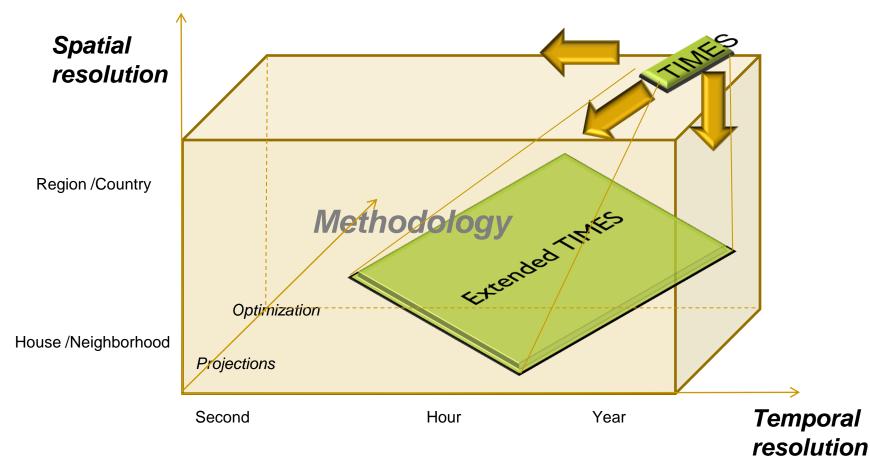


Modeling gaps

Tools have very different scopes, resolution and algorithms.



Research goal





TIMES-MARKAL

TIMES-MARKAL is an energy-economy-environmental model developed under the International Energy Agency's "Energy Technology Systems Analysis Programme".

It is a bottom-up optimization model with the following characteristics:

- It does multi-year optimization (computes the least cost path of an energy system for the specified time frame), but does not have to run every year
- Can be used at the global, multi-regional, national, state/province or community level
- The number and lenght of time slices are defined by the user, within three levels (seasonal (or monthly), weekdays/weekends, hour of the day), with the user being able to choose what degree of resolution to give to each process
- Can test a series of policy options, such as CO2 constraints, taxes or subsidies



TIMES model Inputs/outputs

Supply **Demand** Policy Technoeconomic Drivers Existing Taxes Commodity transformation Subsidies Demand energy Limitation on technologies: curves sources Sectorial Potentials installed electricity Inputs demand Imports capacities generation, Energy Costs services services Availability of consumption Costs consumption resources Efficiencies Storage

Outputs

Installed capacities for each supply and demand technologies, energy fluxes, final energy prices, total system cost, GHG emissions



Modeling dynamics in TIMES

The models being developed are new applications of TIMES, as they try to include some supply and demand dynamics, with higher than usual time resolution.

Each model is divided into 288 time periods of the year:

- 4 seasons
- 3 days per season (Saturday, Sunday and weekday)
- 24h per day ───── Main new feature

Supply dynamics were included in the wind, hydro and geothermal resources, as different periods have different availabilities

11 different sectors for electricity demand, with the domestic sector divided in 9 subsectors. Each sector and subsector has a different load curve for each day.

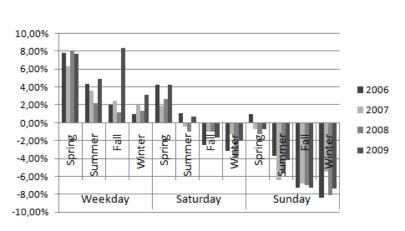
Three models have been built using TIMES: São Miguel, Flores and Portugal (CCS, Waves)



Modeling São Miguel island with TIMES

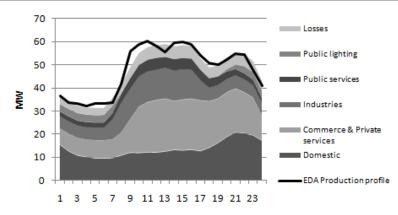
The reconstructed load curves show that the model is able to estimate with some accuracy the evolution of the demand curve through the years.

Some problems still exist in the model as weekdays are usually overestimated and Sundays are underestimated.

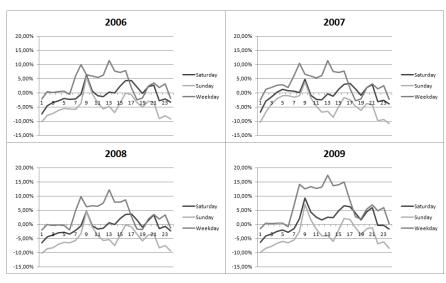


Average relative error for each day





Reconstructed load curve



Average relative error for each hour of each type of day

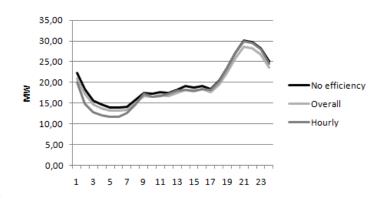
Modeling São Miguel island with TIMES

Modeling the impact of eliminating standby power:

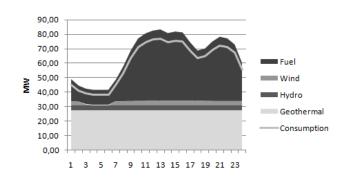
- Reduction of 5% in each hour of domestic consumption (Overall scenario)
- Different hours have different reduction potentials (Hourly scenario)

X.	Wind power installed [MW]	Electricity produced from wind energy [GWh]
Overall	15.27	29.65
Hourly	9.81	18.49

Even with the lower installed capacity, some wind turbines would not be able to produce electricity during the night.



Different scenarios for the domestic curve in 2015



Electricity production for a specific day in 2015, for the Hourly scenario



Modeling Flores island with TIMES

Scenario based approach to study different future energy options:

General efficiency

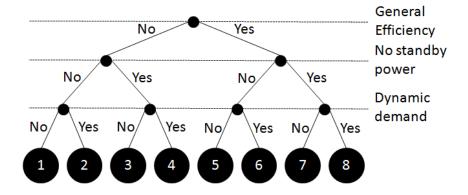
• If there is an increase in overall energy efficiency, demand growth is reduced to 50% of what it would have been using a linear trend

No standby power

 Gradually eliminate stand-by power (starting in 2011 and disappearing completely by 2015). Stand-by power is estimated to account for 5% of the electricity consumed in the domestic sector in Portugal

Dynamic demand

• Gradually enable washers, dryers and dish washing machines to be operated remotely by the grid operator when it is more convenient. Start of introduction in 2013, with all machines having this capability by 2018.

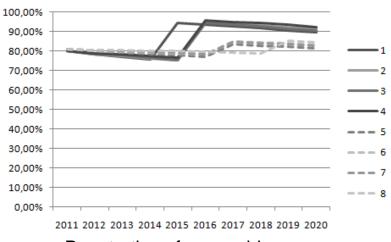




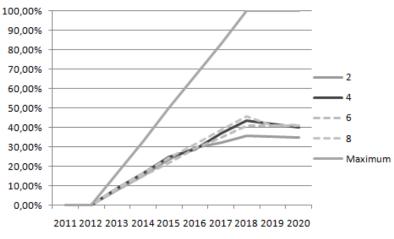
Modeling Flores island with TIMES

Higher demand growths lead to larger investments in renewable energies, thus allowing a higher penetration of renewable energies.

The load shifting capabilities were used to increase the capacity factors of the installed renewables, and postpone the need to install more generation capacity.



Penetration of renewables



Fraction of load shifted



Comparison with other modeling methodologies – Flores case study (Gustavo+Vitor)

Comparison between three methodologies for modeling 1 year:

Dynamic: 8760 hours (ENERGYPLAN)

Semi-Dynamic: 288 time-periods (applied in TIMES)

Integral: 9 time-periods (methodology of LEAP)

Goal: compare the losses on renewable energy production captured by each methodology

Resource	Installed Power 2008 (kW)	Scenario A 2020	Scenario B 2020	Scenario C 2020	Scenario D 2020
Diesel	2,700	2,700	2,700	2,700	2,700
Total Fossil	2,700	2,700	2,700	2,700	2,700
Hydro	1,480	1,480	1,480	1,480	1,480
Wind	600	600	600	600	600
New Hydro	0	0	600	1,200	1,200
New Wind	0	500	1,000	1,500	2,200
Total Renewable	2,080	2,580	3,680	4,780	5,480
TOTAL	4,780	5,280	6,380	7,480	8,180
TOTAL / Peak	2.49	1.61	1.94	2.28	2.49
Renewable / Peak	1.08	0.79	1.12	1.46	1.67



Comparison with other modeling methodologies <u>– Flores case study</u>

Energy Supplied by Renewables

Scenarios	Integral	Dynamic	Semi-dynamic
Scenario A	33.9%	33.9%	34.0%
Scenario B	48.5%	48.0%	48.6%
Scenario C	63.0%	58.9%	62.6%
Scenario D	72.9%	63.9%	71.1%

Energy not used (KWh)

Scenarios	Integral	Dynamic	Energy wasted kWh (Semi-dynamic)
Scenario A	0.00	10	0
Scenario B	0.00	95,977	148
Scenario C	0.00	832,764	111,351
Scenario D	0.00	1,874,367	424,675

> 600 kW wind capacity

Scenarios	Waste in % of Demand (Integral)	Waste in % of Demand (Dynamic)	Waste in % of Demand (Semi- dynamic)
Scenario A	0.00%	0.00%	0.00%
Scenario B	0.00%	0.46%	0.00%
Scenario C	0.00%	4.03%	0.54%
Scenario D	0.00%	9.08%	2.06%

Scenarios	Emissions tCO2 (Integral)	Emissions tCO2 (Dynamic)	Emissions tCO2 (Semi-dynamic)
Scenario A	3,411	3,411	3,406
Scenario B	2,661	2,685	2,654
Scenario C	1,911	2,119	1,930
Scenario D	1,397	1,866	1,491



Methodological thinking

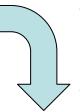
In order to reduce the computational complexity of the problem, the proposed methodology consists in the use of two different tools.

Medium-term model

- Multi-year optimization of investments in renewable energies
- Some hourly dynamics
- Detailed description of energy consumption:
- Across different sectors
- Different types of energy carriers
- •Capable of understanding the evolution of the system:
- Economic growth
- Fuel prices
- Energy demand

Introduction of restrictions for optimization of investments





Test feasibility of the results, one year at a time

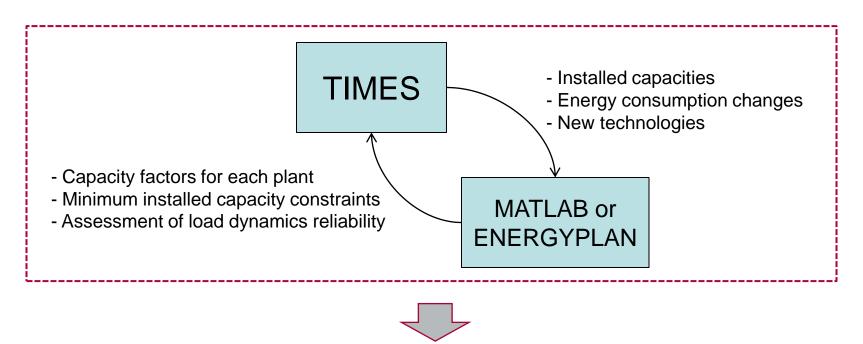
Short-term model

- Hourly optimization of electricity production for one year, taking into account:
- •Start-up costs and efficiencies for thermal engines
- Variability of renewable resources
- Solar
- Wind
- Hydro
- Fuel prices
- Demand profiles
- Dynamic demand options (EVs, others)
- Optimization of use of energy storage systems



Methodological thinking

TIMES (4 seasons, 3 days, 24h) + Short-time model for key years



- Scenarios for capacities that can be installed (including storage technologies)
- More robust results regarding power system operation reliability and security



Short-term model for S. Miguel

MATLAB model being developed by Gonçalo Pereira (MSc)

Optimization outputs Inputs Data generation Installed capacities for Electricity demand curve Electricity production by electricity generation by for the whole year source source Hourly availability for each Charge and discharge of Energy storage size and renewable resource energy storage unit capacity For wind energy, Overall electricity demang historical hourly capacity factors for other islands growth are used, São Miguel-8760s Aug Sep Oct



The proposed methodology was tested using the São Miguel TIMES model.

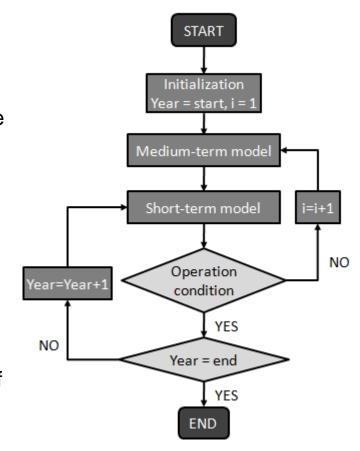
The model had to make two decisions:

- When should the 2 x 10 MW Geothermal facilities be installed
- What amount of wind energy should it install and when

The methodology was applied separately for the two decisions.

Operation conditions:

- Geothermal: the plant must have a capacity factor of 90% or higher for at least 95% of the time.
- Wind: the wind turbines produce at least 90% of the nominal capacity factor.





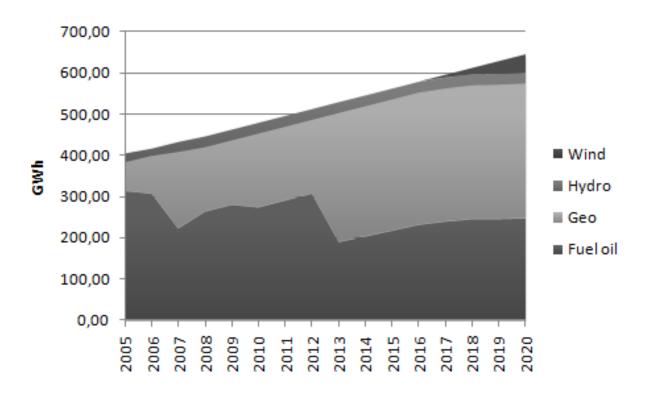
Total installed capacity of wind energy after each TIMES iteration. Some notes:

- TIMES processing is the first iteration of TIMES, without any constraint.
- Iteration 1 is the last iteration of the Geothermal decision process, and the first of the Wind decision process.
- Iteration 10 was the last iteration of the Wind decision process.

Iteration	2011	2012	2013	2014	2015	2016	2017	2018	2019	2020
TIMES processing	0.0	0.0	0.0	0.0	0.0	0.0	3.5	8.3	16.3	23.3
1	9.7	28.7	28.7	28.7	28.7	30.5	30.5	30.5	30.5	30.5
2	9.7	12.2	26.3	26.3	27.3	29.9	29.9	29.9	29.9	29.9
3	9.7	12.2	15.4	16.9	27.3	29.9	29.9	29.9	29.9	29.9
4	9.7	9.9	9.9	9.9	27.3	29.9	29.9	29.9	29.9	29.9
5	9.7	9.9	9.9	9.9	9.9	29.9	29.9	29.9	29.9	29.9
6	9.7	9.9	9.9	9.9	9.9	9.9	28.7	28.7	28.7	28.7
7	9.7	9.9	9.9	9.9	9.9	9.9	13.5	27.0	27.0	27.0
8	9.7	9.9	9.9	9.9	9.9	9.9	13.5	14.8	14.8	22.3
9	9.7	9.9	9.9	9.9	9.9	9.9	9.9	9.9	9.9	22.3
10	9.7	9.9	9.9	9.9	9.9	9.9	9.9	9.9	9.9	9.9

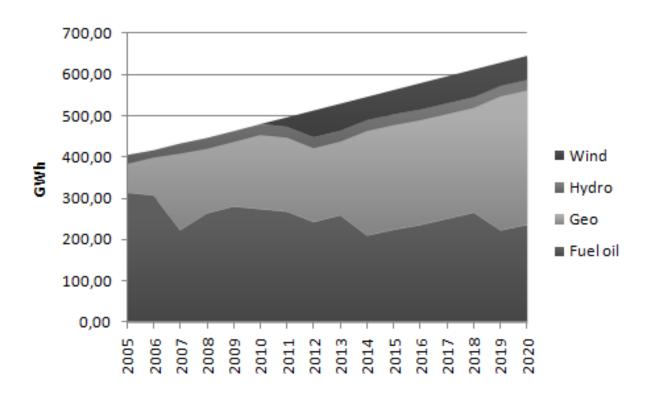


Results for the iteration TIMES processing



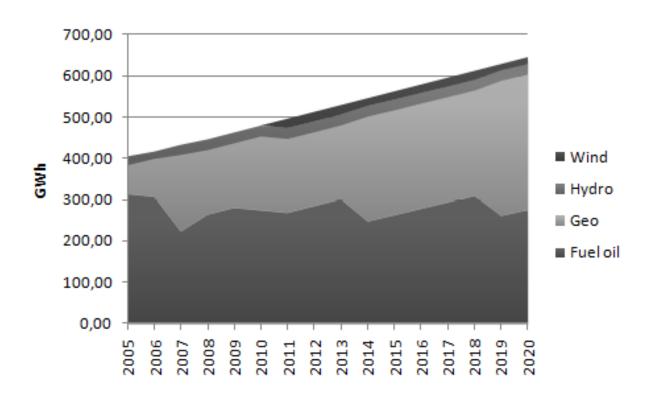


Results for iteration 1



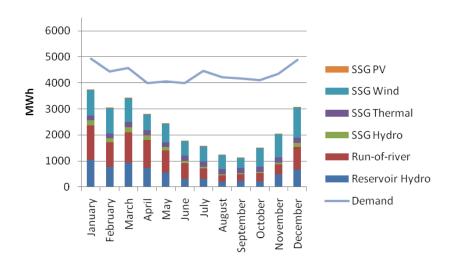


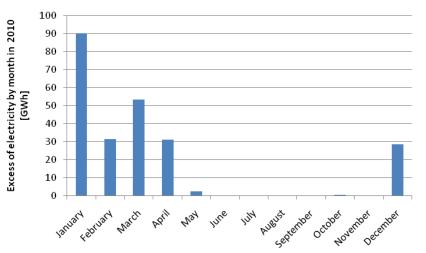
Results for iteration 10





- In 2010, renewables produced
 ~50% of all the electricity in
 Portugal
- Some periods during the Winter time had excesses of renewable electricity
- The investment in renewable generation capacity should be analysed with high temporal resolution







MW	2005	2006	2007	2008	2009	2010
Coal	1776	1776	1776	1776	1756	1756
Oil	1909	1909	1877	1877	1878	1822
Natural Gas	2166	2166	2166	2166	2992	3829
Large Hydro	4582	4582	4578	4578	4578	4578
Thermal Status Producers	1166	1295	1365	1424	1610	1698
Hydro Status Producers	333	365	374	385	395	410
Wind Status Producers	891	1515	2048	2662	3357	3705
Solar Status Producers	0	0	13	53	95	122
Wave Status Producers	0	0	0	2	2	2

MW	2011	2012	2013	2014	2015	2016	2017	2018	2019	2020
Large Hydro	4524	5231	5231	5476	6467	7489	8394	8712	8798	8798
Small Hydro	457	503	503	550	550	600	650	650	700	750
Wind onshore	4928	5600	5600	5600	6100	6100	6100	6600	6800	6800
Wind offshore	0	0	0	0	25	25	25	25	25	75
Solar	258	340	465	590	720	860	1005	1160	1325	1500
Wave	5	5	10	35	60	75	100	125	175	250

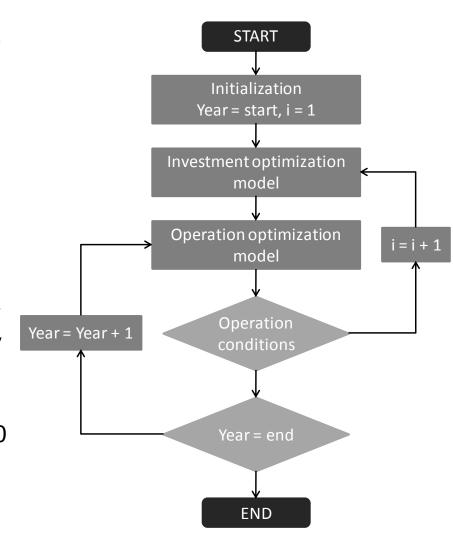


Application of methodology with TIMES and EnergyPLAN.

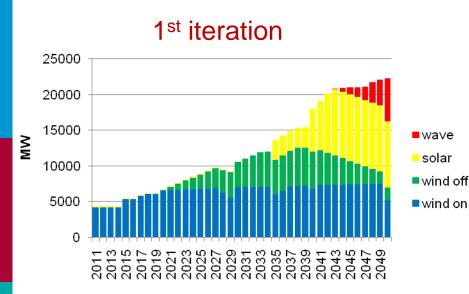
Time horizon of 2005-2050.

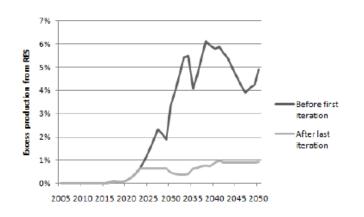
Mainland Portugal case study

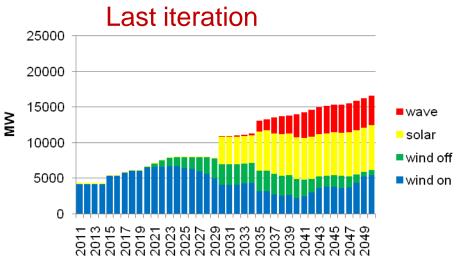
- Operation conditions:
 - Maximization of renewable energy penetration such that the last installed MW produces at least 90% of its potential capacity factor
- 2 scenarios concerning installed capacity for hydro pump storage (Current capacity of 1036 vs expected capacity of 4302)
- Assumed gradual reduction of CO2 emissions to ~30% of 2005 levels in 2050



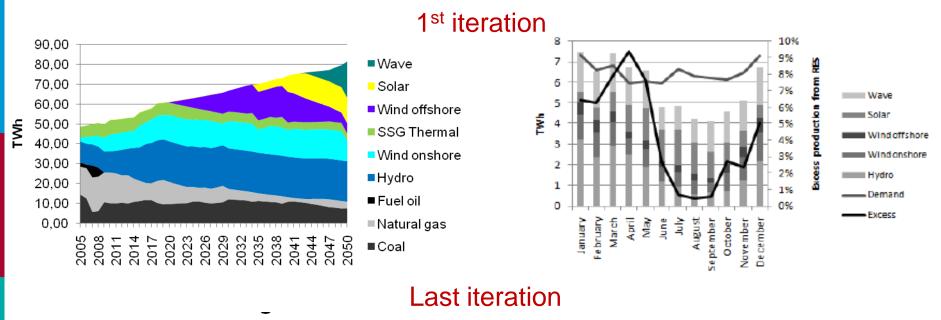




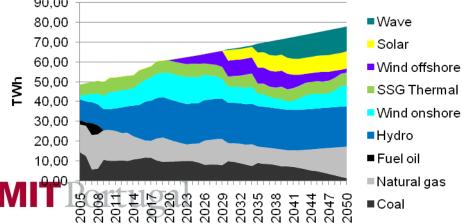


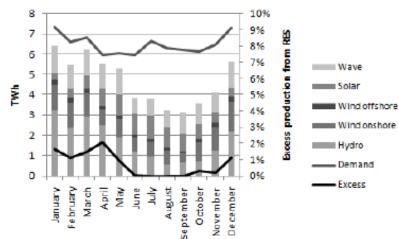






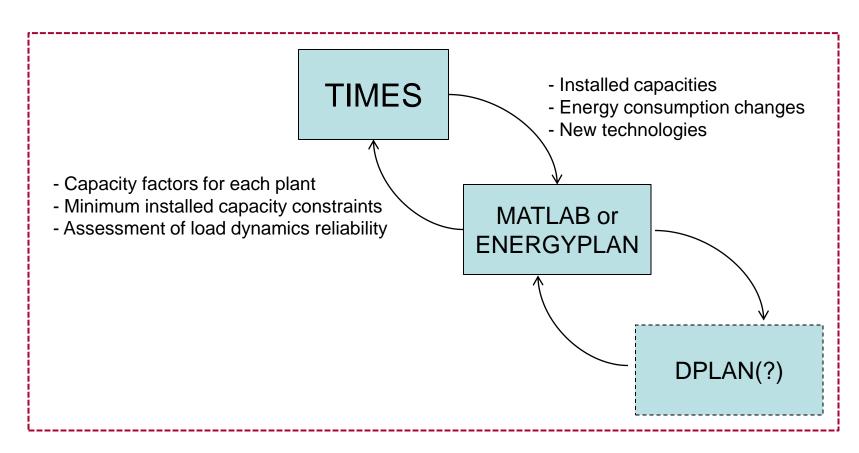






Next Methodological Step

TIMES (4 seasons, 3 days, 24h) + Short-time model for key years

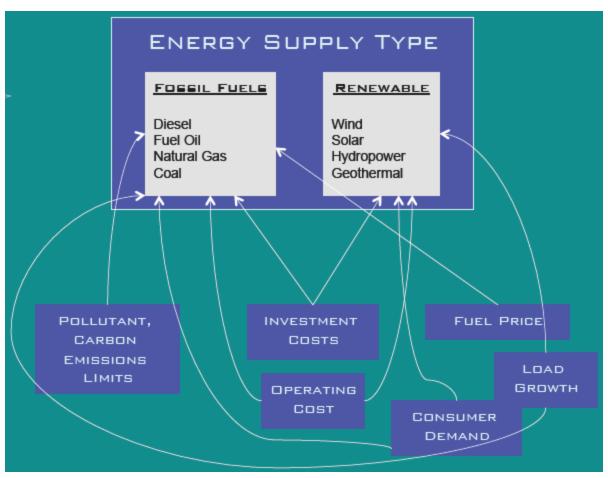




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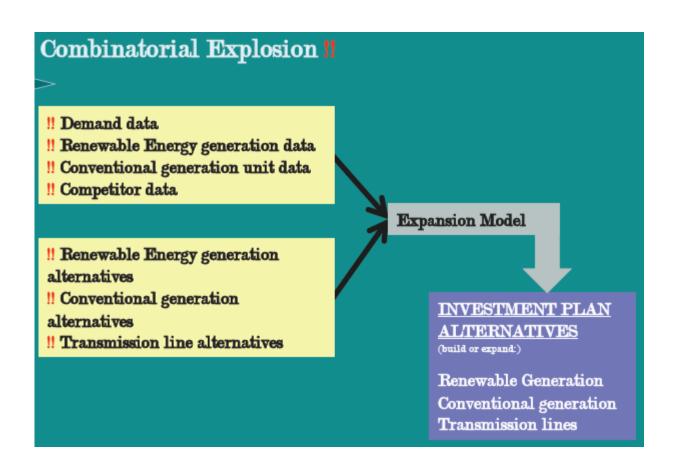
Other ideas going on:

Flexible Networked System Design Under Uncertainty: A Case Study in Long-Term Energy Planning



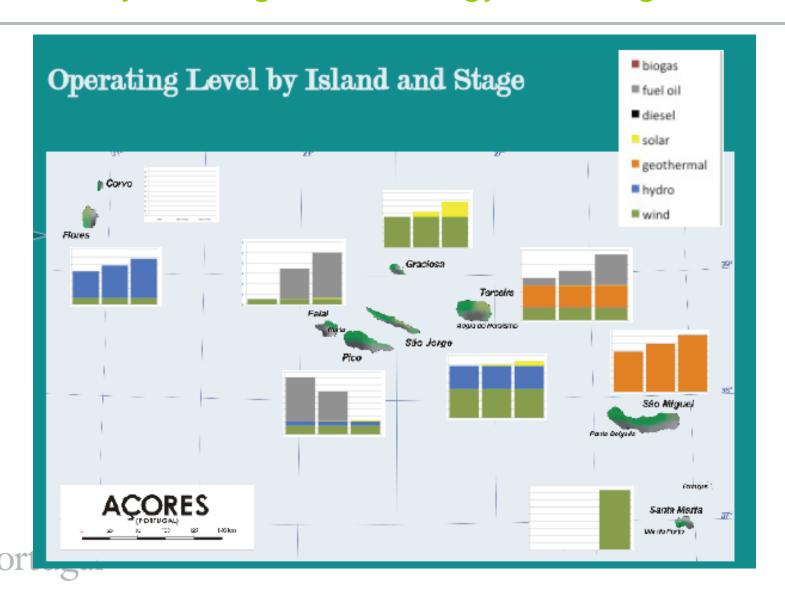


Flexible Networked System Design Under Uncertainty: A Case Study in Long-Term Energy Planning

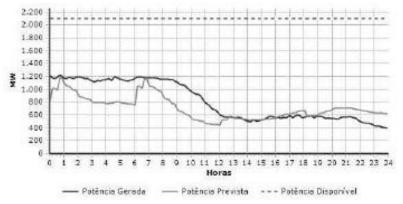




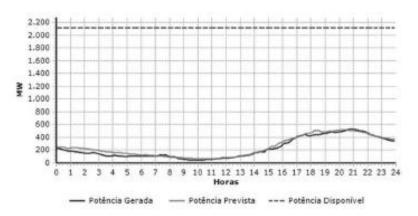
Flexible Networked System Design Under Uncertainty: A Case Study in Long-Term Energy Planning



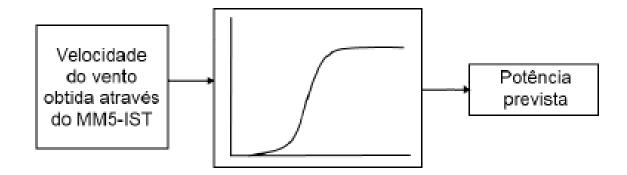
Wind forecasting using sof-computing





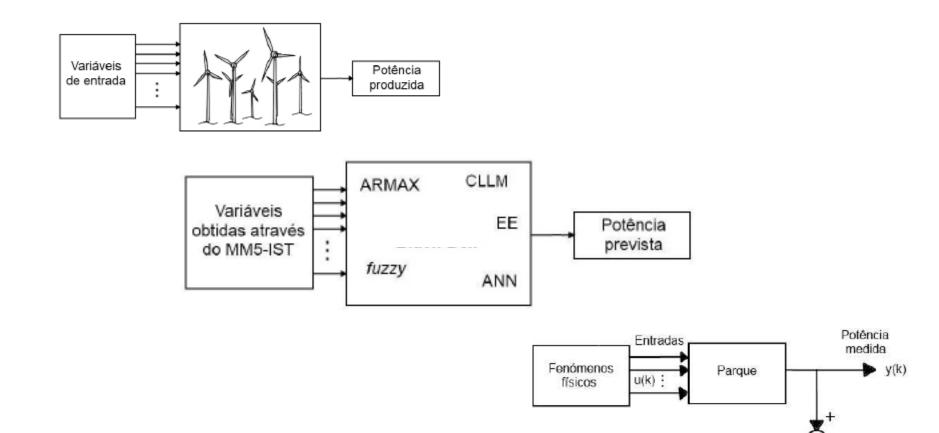


(b) 1 de Julho de 2010.





Wind forecasting using soft-computing



Potência

modelada

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Entradas

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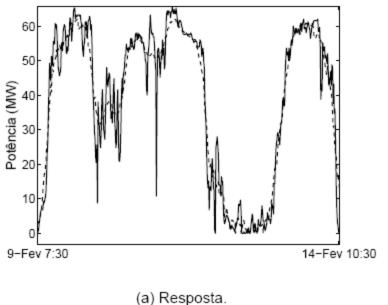
Mozelo

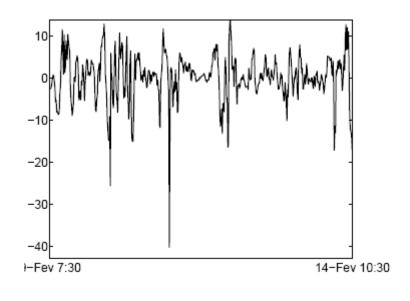
Modelo

numérico



Wind forecasting using soft-computing





(b) Resíduo.

