Long-term assessment for sustainable evolution of cities: combining land-use and building allocation with an energy system model

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Context

Population (Millions)	GHG Emissions (M tCO ₂ e)	GDP (billion \$ PPP)
1. China: 1,192	1. USA: 7,107	1. USA: 14,204
2. India: 916	2. China: 4,058	2. 50 Largest Cities: 9,564
3. 50 Largest Cities: 500	3. 50 Largest Cities: 2,606	3. C40 Cities: 8,781
4. C40 Cities: 393	4. C40 Cities: 2,364	4. China: 7,903
5. USA: 301	5. Russian Federation: 2,193	5. Japan: 4,354
6. Indonesia: 190	6. Japan: 1,374	6. Top 10 GHG Cities: 4,313
7. Brazil: 159	7. Top 10 GHG Cities: 1,367	7. India: 3,388
8. Russian Federation: 142	8. India: 1,214	8. Germany: 2,925
9. Top 10 GHG Cities: 136	9. Germany: 956	9. Russian Federation: 2,288
10. Japan: 128	10. Canada: 747	10. United Kingdom: 2,176

Source: See Annex D. Data for the urban agglomeration associated with each C40 city is used in calculations to maintain consistency with the 50 largest cities, 2005.

Figure: The 50 largest cities, C40 Cities and 10 GHG emitting cities position amongst population, GHG emissions and GDP top countries.

source: Worldbank, 2012



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Figure: Percentage of population in urban and rural areas in France.

source: United Nations, Department of Economic and Social Affairs, Population Division (2018).



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Energy coupling and perspectives

Urban planning revolution in France





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Addressing	urban	growth
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Energy coupling and perspectives

What can prospective modeling bring to the issue?

Need for:

- Long-term study;
- Scenario comparison;
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Optimization bottom-up models (*TIMES* from IEA-ETSAP):

- Scenario-based approach;
- **Territory** centered (no general solution);
- **Optimization** flexible tool;
- Integrating diverse sources of information.



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Data gathering



Figure: Bordeaux Metropolitan territory, colored by cities (27 gathered) and separated by IRIS zones.

From Bonhomme, 2013, Adam et al., 2005, Cerezo et al., 2017, Gonçalves et al., 2017 and Delmastro et al., 2016 among others.

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Data processing: PCA/ k-means

Parameter	Contribution to first axis
Floor area ratio	-0.258866226
Average size of lots (log m ²)	0.256179325
Average coverage ratio of lots	-0.247272853
Average blocks area (log m ²)	0.243078477
Average coverage ratio of non empty lots	-0.241760861
Average distance to public transportation (log m)	0.239797541
Lots' density	-0.239673269
Ratio of road surface	-0.237407088
Ratio of road length	-0.234629399
Building coverage ratio	-0.232312303
Ratio of non-empty lots	-0.22799179
Teaching equipment (per m ²)	-0.22406754
Health equipment (per m ²)	-0.221496668
Lots' width average	0.218652122
Lots' height average	0.213608755
Entropy coefficient from Corine Land Cover data	0.212987077
Ratio of vegetalized area	0.204970888
Average size of lots (m ²)	0.162629794



Table: List of parameters ordered by contribution to PCA's most important axis.



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Name	Description	Nb IRIS	Teaching equipment (per m ²)	Area of green spaces	Block area (log m ²)	Ratio of vegetalized area
Dense close city 1	Bordeaux's highly dense center (small lots, few green spaces, few variety of land cover, high CES, much road sur- face)	34	0.02	1708.18	8.83	0.08
Suburbs / smaller towns 1	Similar lots of suburbs/smaller towns (average CES, small lots, few road sur- face)	63	1.4E-4	3668.88	9.45	0.21
Rural area 1	Rural areas (few schools, larger portion of green areas, diverse land-use, very little built or road surface, very large lots)	16	1.63E-06	20931.12	11.15	0.51
Historical centre	Highly dense historical center (small lots, few green spaces, high CES, much road surface, many schools/health in- frastructure)	13	0.0038	671.66	8.28	0.03
Dense close city 2	Bordeaux's dense diverse further sur- roundings (variety of land-use, fewer public equipments, smaller CES, rather dense and small lots)	45	3.3E-4	1709.28	9.25	0.11
Suburbs / smaller towns 2	Similar lots of peri-urban small towns/suburbs (average/small CES, few equipments, normal/large lots, few road surface)	43	2.91E-05	7824.59	9.97	0.33
Activity zone 1	Activity zones around the metropolitan territory	4	2.91E-06	2.51E+04	1.11E+01	3.12E-01



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Outputs

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Figure: Map of clusters created on parameters concerning land-use.



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Reference land system





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Energy related and environmental outputs

Urban constraints:

- Limited areas by local plans;
- Historical center and other specific areas;
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Energy related and environmental outputs

Urban constraints:

- Limited areas by local plans;
- Historical center and other specific areas;
- Lifespan of buildings;

Environmental constraints:

- WHO recommendations on *green areas access* [WHO Regional Office for Europe, 2016];
- Limited artificialization of the land;
- Retrofitting program to diminish buildings performing weakly;



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Results: land-use and buildings



500 400

300

200

100

0

-100-200

Base

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Results: land-use and buildings



Activity buildings 1 (2000-2012)

High 1970-1980 housing Historical buildings (< 1915) Historical buildings (< 1915)

Middle size 1 (1970-1980) Middle size 1 (1970-1980) Middle size 2

Small housing suburbs 1 Small housing suburbs 1 Small housing suburbs 2 (1970-2000)

Small housing suburbs 3 Small housing suburbs 3

Activity buildings 2

Activity buildings 2 High 1970-1980 housing

Middle size 2 Rural housing Rural housing

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Results: DPE constraint



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Results: DPE constraint







How to feed the model

Land-use optimization model

Buildings

Energy coupling and perspectives

Results: DPE constraint









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Energy model scheme



Figure: Reference Energy System



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Perspectives			

Three approaches for the **coupling**:

- Changing **final demand** using converting tools based on land-use model outputs [Le Gallic et al., 2016];
- Changing **technologies** themselves proposed to the model;
- **Constraining potential** of each technology (e.g. no collective heating if no collective buildings are built).



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Other perspectives on land-use:

- Add urban planners' future projects;
- Combine with jobs and commercial sector data;
- Link transportation demand with urban shapes;



Thank you!

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