

Influence of weather sensitivity on electricity consumption in Abidjan, Cotonou and Lomé, three Coastal Cities in Western Africa

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Background and Context

Data and Analysis Methods (2 | 2)

Results : Weather & consumption indexes (1 | 2)

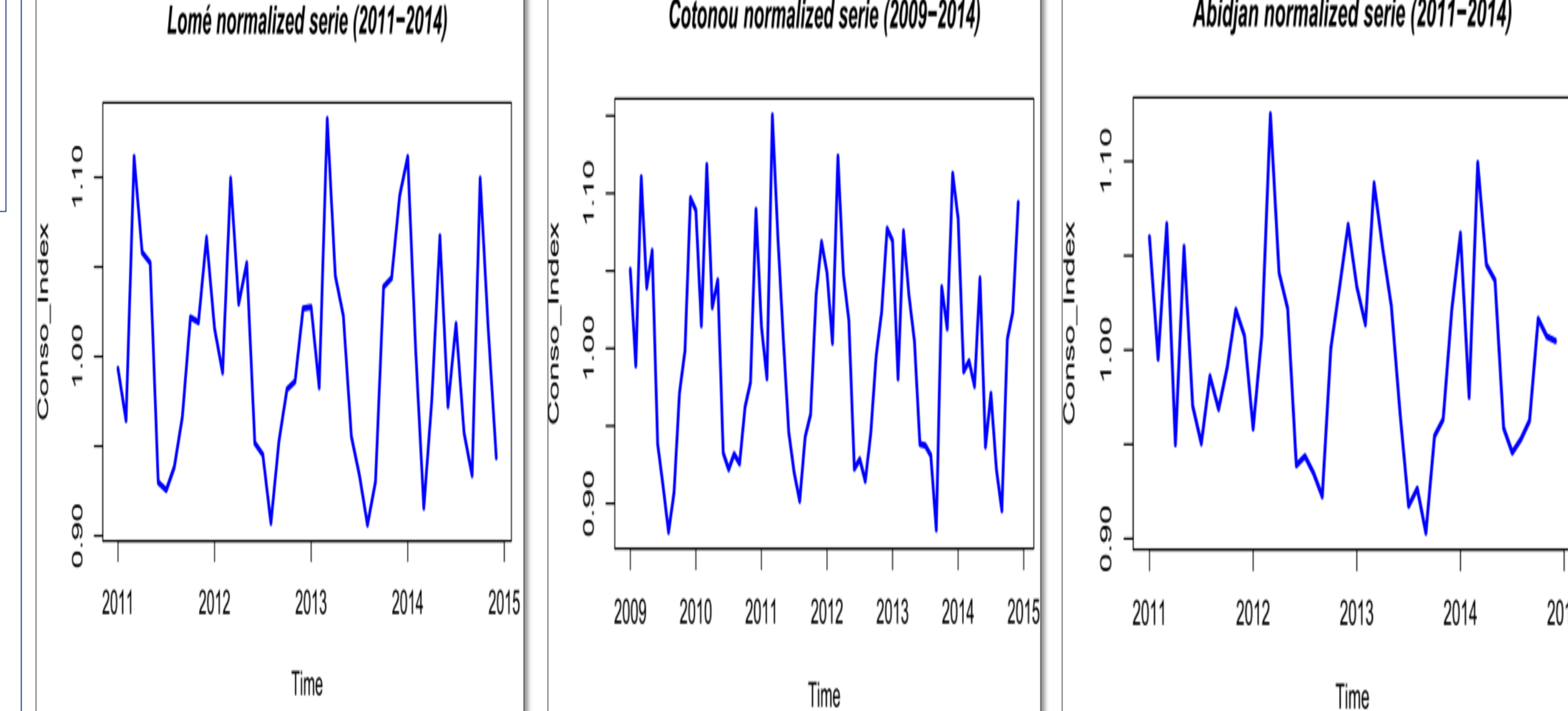
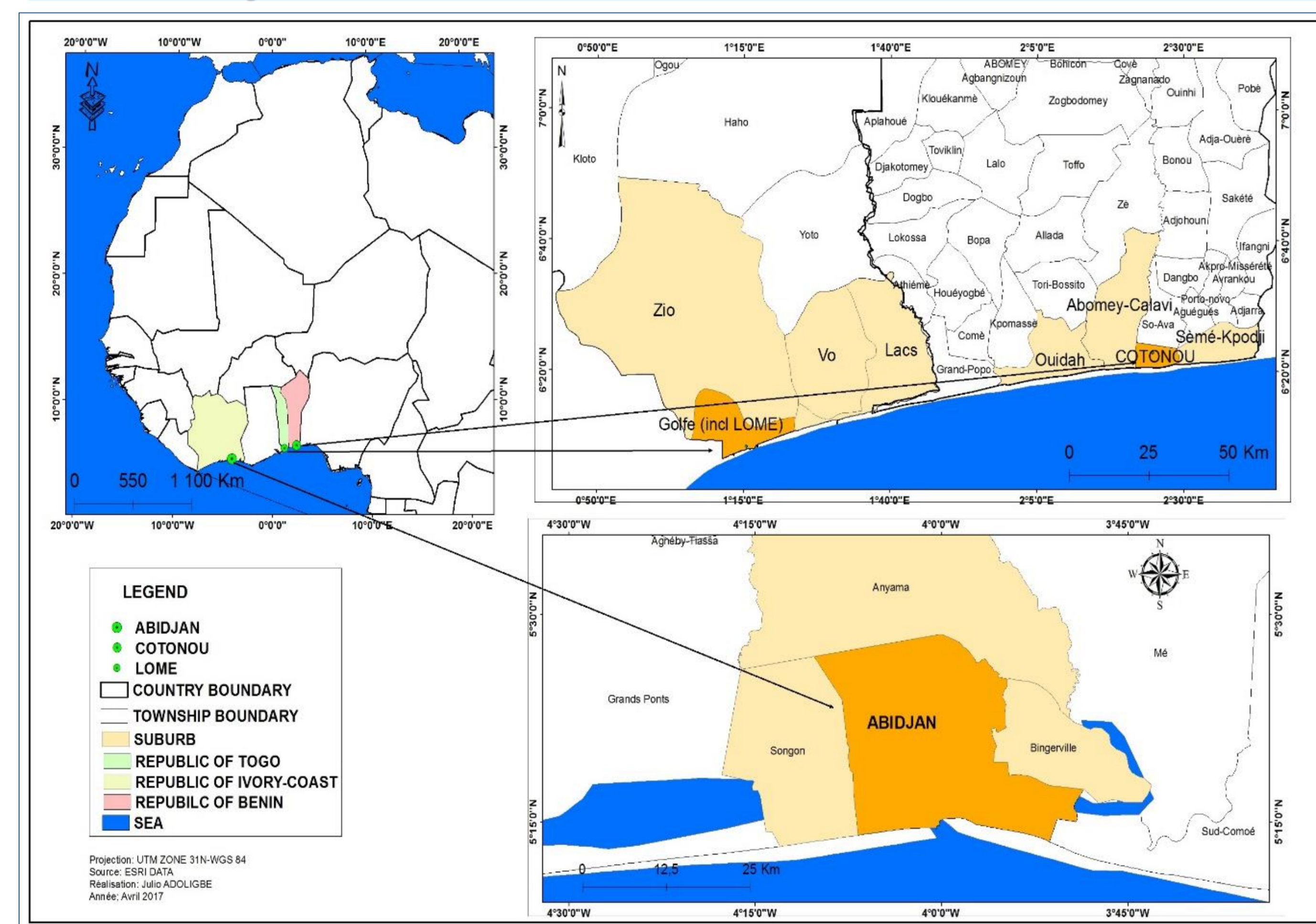
References

- ❖ Low carbon economy requires massive development of VRE which depend on weather.
- ❖ Power system sensitivity to weather variability (Hence the electricity consumption (E.C) too).
- ❖ From published of regions worldwide, electricity consumption can depend on air temperature, relative humidity, wind speed or daylight time.
- ❖ Need for understanding how electricity consumption is related to weather variables in the context of Climate change.
- ❖ The present work, explore how weather variability has driven the electricity consumption of western Africa's coastal cities in the last years using Cooling-Degree-Days (CDD).

- ❖ **Time evolution of monthly consumption is at first governed by non-climatic factors : GDP, Urbanization, population growth etc**
- ❖ Selection of homogeneous periods (2011 - 2014 for all cities), periods with no political / economical concerns
- ❖ Normalization : remove consumption drift likely related to GDP(trend)
- ❖ Normalization with a classical Multiplicative Decomposition of the time series.

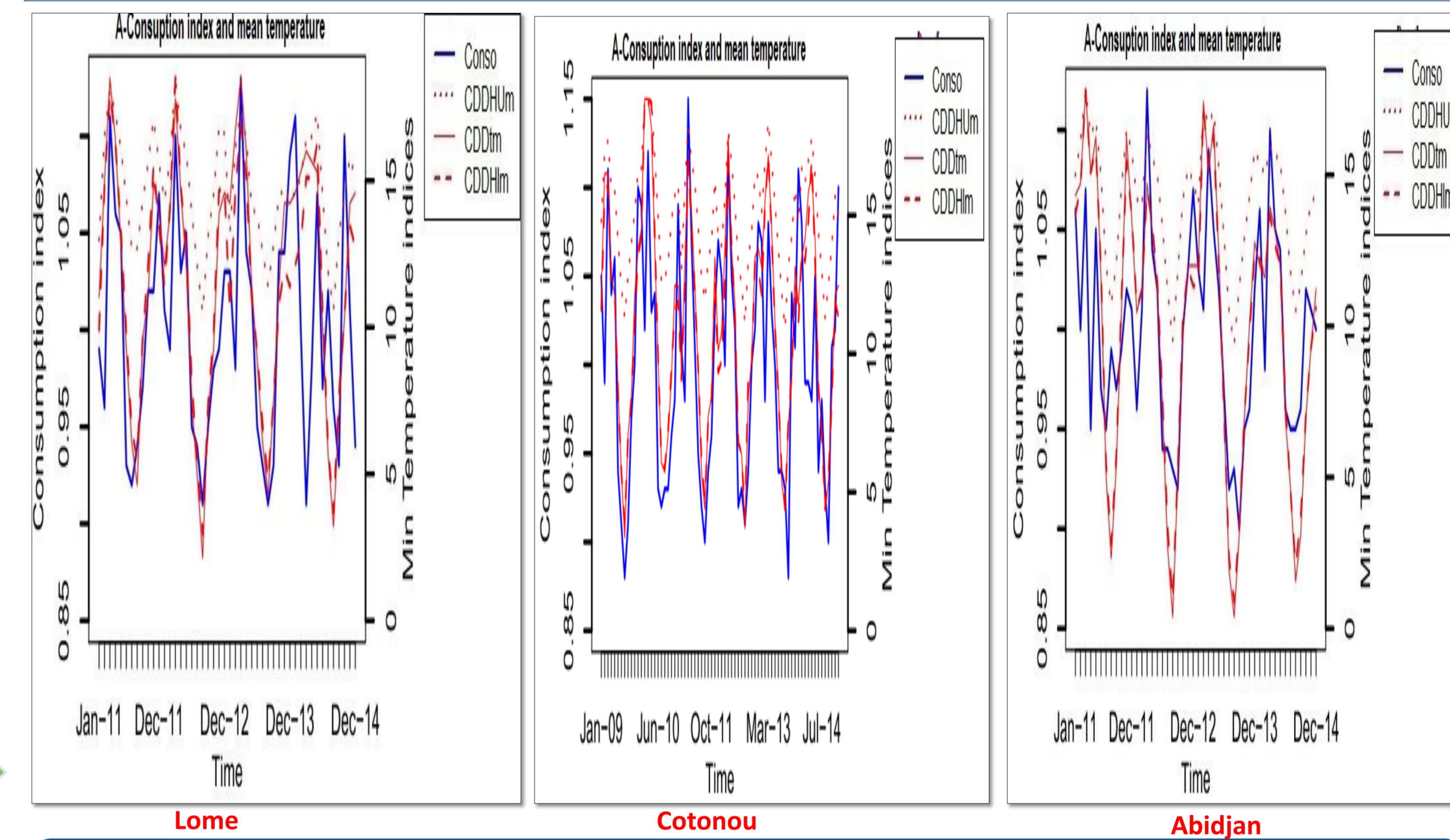
- ❖ **Inter-annual evolution of monthly CDDs and of monthly electricity consumption**

Study case: Tree coastal cities in West Africa



Normalization over the working period

- ❖ High seasonality : (1)a peak every March (warm period; warm and dry), (2) a trough every August ("little dry season"; warm and humid)
- ❖ **Appropriate National Weather Station data is a challenging too**
- ❖ seasonality : (1)a peak every March (warm period; warm and dry), (2) a trough every August ("little dry season"; warm)
- ❖ **Raw Weather data**: daily dew point; temperature , relative humidity from 1980 to 2014 from the national company of the three cities



Relation between consumption and cooling degree day: an additional consumption due to buildings cooling

Pearson correlation (p-value > 0.05.) between monthly consumption (Conso_index) and temperature index CDDs .

CORRELATION MATRIX							
	Conso_index	CDDtm	CDDHUm	CDDHIm	CDDTxn	CDDHUxn	CDDHIxn
Abidjan							
conso_index	1.00	0.76	0.75	0.75	0.76	0.75	0.75
Cotonou							
conso_index	1.00	0.78	0.73	0.75	0.78	0.73	0.78
Lomé							
conso_index	1.00	0.66	0.59	0.63	0.66	0.59	0.63

Better links in Abidjan and Cotonou

Conclusion & Perspectives

For the three (3) cities :

- ❖ Strong relationship between temperature with increased consumption in hot weather conditions
- ❖ Influence of relative humidity for Cotonou and Abidjan
- ❖ Heat index and temperature : better explanatory power than

Further investigations

Electricity demand Model

Approach similar to Scapin et al., 2016 & Waite et al., 2017

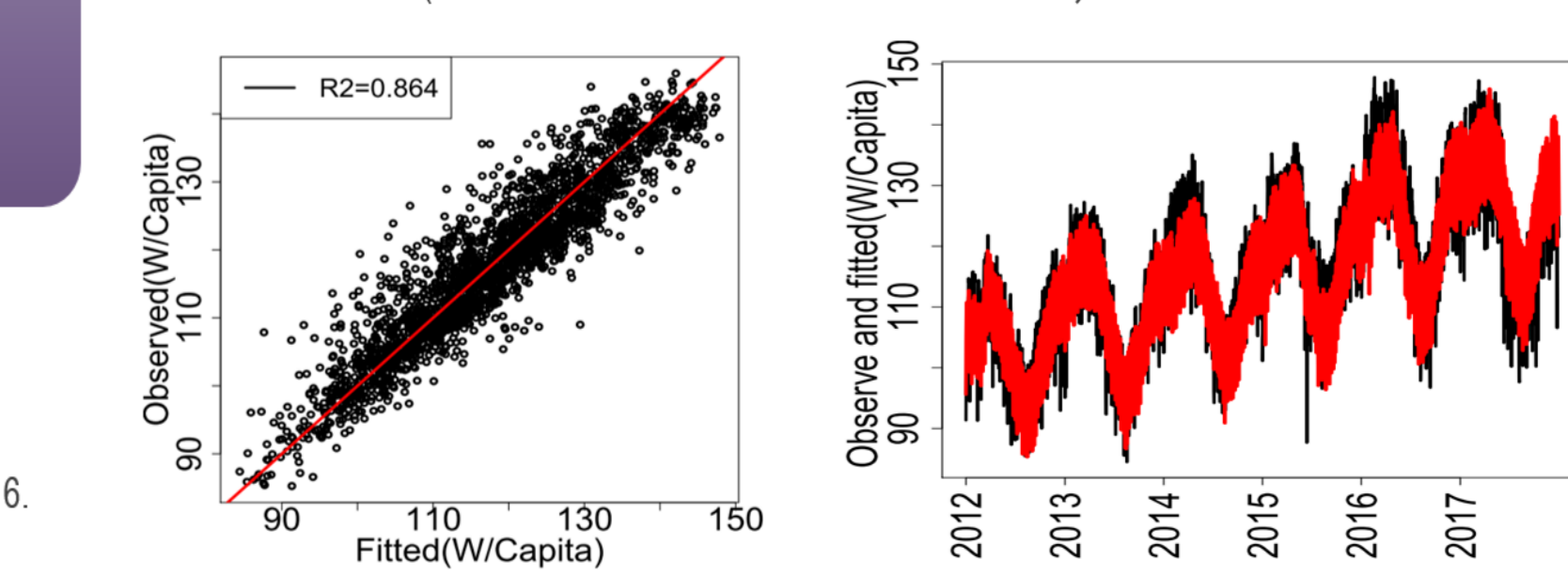
$$C(t) = \left\{ C_0 + \beta_0.IT(t) + \beta_1.IRH(t) + \sum_{i=1}^7 \gamma_i.IJ(i) + \varepsilon(t) \right\} + f(t)$$

With
 C(t) : per capita consumption for day t.
 f(t) : temporal trend
 C0(t) : basis consumption [W/Capita].
 β_0 (t) : temperature-sensitivity [W°C/Capita].
 β_1 (t) : humidity-sensitivity [W%/Capita].
 ε (t) : residuals

With
 γ_i (i=7) : daily factors [W/Capita].
 IT(t) : Cooling Degree Days (CDD, °C). IRH(t) : Humidity Index [%].
 IJ(i) : weekday dummies with IJ = 1, 2, 3, 4, 5, 6, 7 for Mondays, Tuesdays, ... & Sundays and holidays together.
 ε (t) : residuals

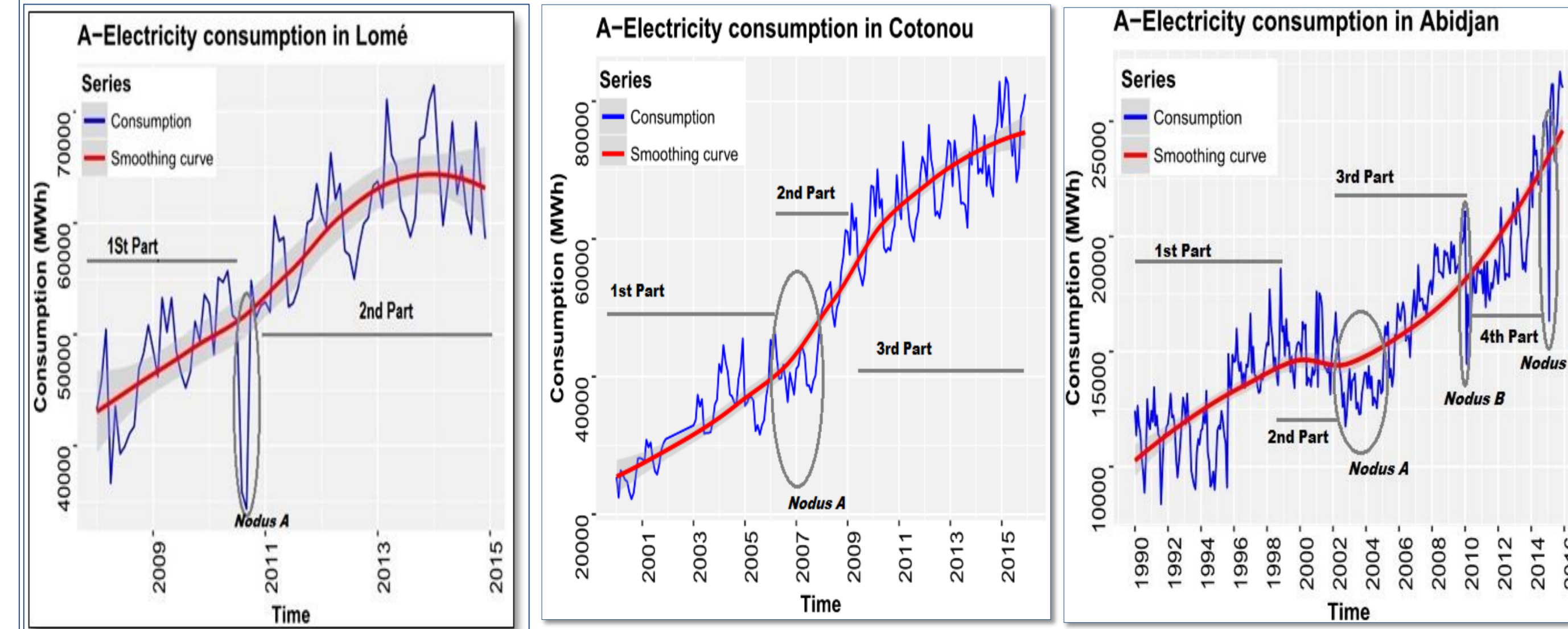
Results : drivers & weather sensitivity

Well-fitted model (correlation between observations and forecasts)

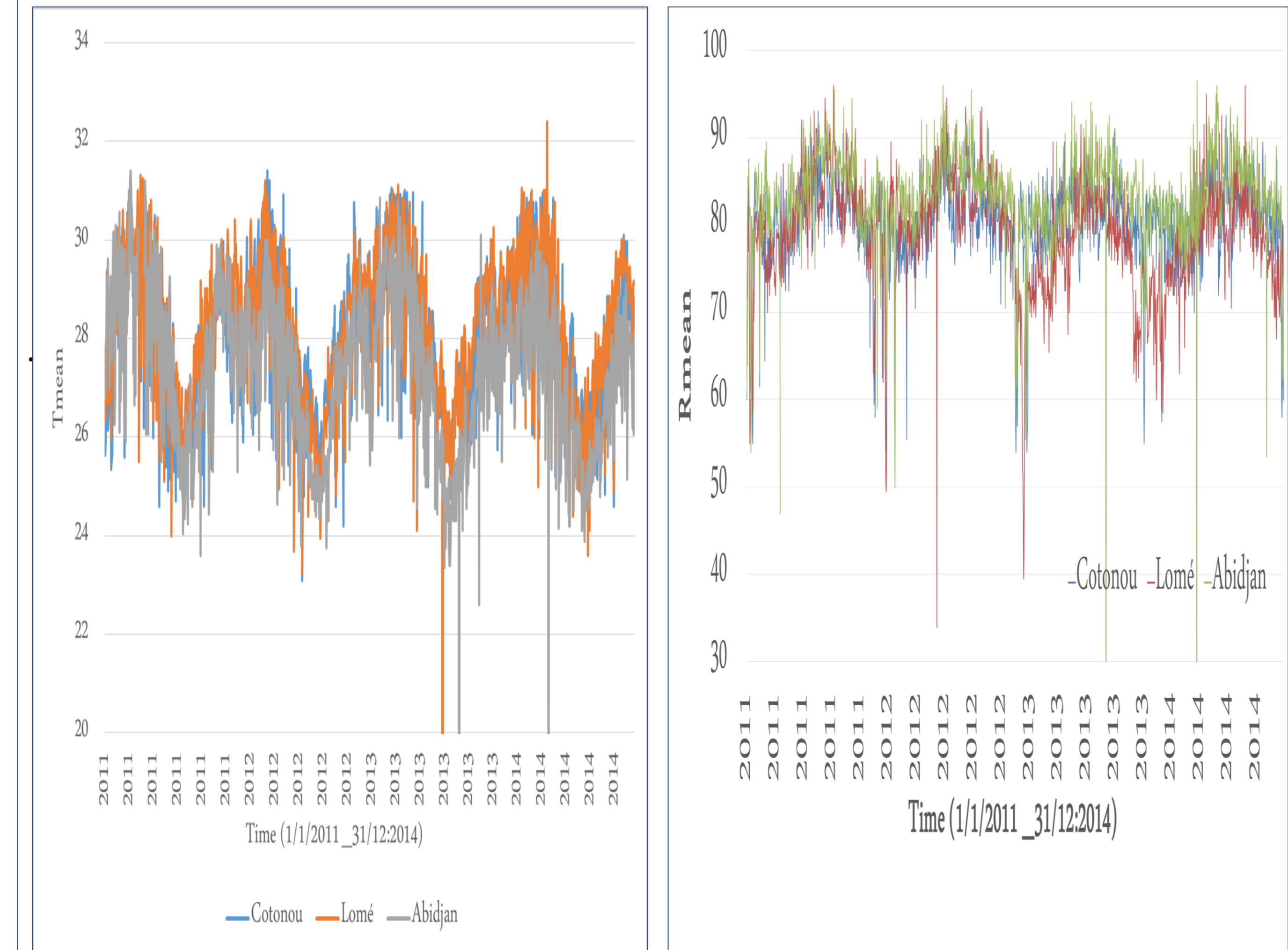


Data and Analysis Methods (1 | 2)

- ❖ **Electricity consumption (raw data : monthly means)**
- ❖ **Getting energy data is a challenge**



Raw data over the whole period



- ❖ **Weather indexes computing**
- Humidex (HU) & Heat Index (HI), functions of temperature and Humidity
- Cooling Degree Days CDD(t) = a (T(t)-Tthreshold)

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