



### Using cloud fraction derived from all-sky camera to improve beam solar irradiance forecasting by time-series modelling

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### **Reuniwatt SAS**

- Young company based in Reunion Island (Indian Ocean) with a Parisian office
- Coupling renewables and IT



### Solar energy and derived forecasting issues

High variability of the solar resource must be monitored to increase its use.

Challenges of solar energy forecasting:

- Converting an intermittent solar energy production into a guaranteed energy
- Massive and safe penetration of PV into the energy mix



Short-term solar energy forecasting applications:

- Smart-grids (network security)
- Storage management
- Trading



### Outline

# 1) Forecasting methods 2) Objective & approach 3) Data used 4) Modelling 5) Simulations & Results 6) Conclusion & Perspectives



### **Forecasting methods**

#### 1) Forecasting Methods

- 2) Objective & approach
- 3) Data used
- 4) Modelling
- 5) Simulations & Results

Different forecasting methods are used for specific forecasting time-horizons

NWP (Numerical Weather Prediction) Day-ahead forecasts Res: ~10 km<sup>2</sup>



Satellite images Hour-ahead forecasts Res: ~1 km<sup>2</sup>



Time-series modelling (ARMA, ANN...) Intra-hour forecasts → Adapted for local forecasting



## Forecasting methods (2)

#### 1) Forecasting Methods

- 2) Objective & approach
- 3) Data used
- 4) Modelling
- 5) Simulations & Results

#### • <u>Since 2011</u>:

**Total sky imagers** are more and more used for nowcasting and deterministic shortcasting (up to 30 min ahead forecasts)

(Gauchet *et al.*, 2012; Chow *et al.*, 2013; Coimbra and Marquez 2013) → Provide pertinent sky information over a single site

#### • Our work:

Preliminary study that uses **sky observations** as **an additional information** to a standard **statistical model** for direct solar irradiance shortcasting (t+10 min, t+20min, t+30 min)







### Approach

- 1) Forecasting Methods
- 2) Objective & approach
- 3) Data used
- 4) Modelling
- 5) Simulations & Results
- Extract cloud cover from a Total Sky Imager



 Use this information as an external input of an ARMAX (Auto-Regressive Moving Average with eXternal Input) model.
 — Reuniwalt —

### Data Used

#### 3 types of data:

- Processed total sky images taken every 10 min from a TSI (*Total Sky Imager*)
- 2) DNI (*Direct Normal Irradiance*) measurements synchronized with images
- 3) Modelled DNI under McClear clear sky model (Lefevre *et al.* 2013) for direct irradiance normalization

- Forecasting Methods
  Objective & approach
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Data extracted from the NREL (*National Renewable Energy Laboratory*) database

NATIONAL RENEWABLE ENERGY LABORATORY



**Data period**: from 01/01/2011 to 30/12/2011 $\rightarrow$  10 819 observations used after filtering (night data, missing consecutive values or images...)

Public data → reproducible results



# Modelling (1): ARMAX model

1) Forecasting Methods

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- 2) Objective & approach
- 3) Data used
- 4) Modelling
- 5) Simulations & Results
- ARMA(X) Model (Auto-Regressive Moving Average with eXternal Input)

$$Y_t = \theta_0 + \sum_{i=1}^{p} \theta_i Y_{t-i} + \sum_{j=1}^{q} \varphi_j \varepsilon_{t-j} + \sum_{k=1}^{r} \beta_k X_{t-k} + \varepsilon_t$$

Where  $\{\varepsilon_t\}$  is a series of independent and identically distributed error terms.  $\rightarrow$  Simple but efficient model

• External Input = cloud cover index:

$$X = i_{cs} = \frac{number \ of \ pixels \ corresponding \ to \ a \ cloud}{number \ of \ pixels \ of \ the \ sky \ part}$$

### Modelling (2) : Parameter Estimation

1) Forecasting Methods

- 2) Objective & approach
- 3) Data used
- 4) Modelling
- 5) Simulations & Results

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#### **Parameter Estimation**

• Parameters to be estimated:

 $\boldsymbol{\Theta} = (\theta_0, \dots, \theta_p, \varphi_1, \dots, \varphi_q, \beta_1, \dots, \beta_r)$ 

- Estimation by **RLS** (*Recursive Least Squares*) method:
  - Parameters updated after each new observation  $\Theta \rightarrow \Theta_t$
  - Same weight for each passed observation
  - Algorithm well adapted for operational forecasting (low complexity)
- Parameters  $\Theta_t$  calibrated with half the data. Second half used for testing.

#### Choice of the model

- Order *p*, *q*, *r* of ARMA and ARMAX models chosen by grid-search
- One model per forecasting horizon (t+10min, t+20min, t+30min)

# Simulations & Results

- Forecasting Methods
  Objective & approach
  Data used
  Modelling
  Simulations & Results
- All simulations made on DNI normalized by McClear Direct Irradiance.
- ARMA and ARMAX models are compared with persistence model.
- Criterion of comparison (on non-normalized DNI):

RMSE(h) = 
$$\sqrt{\frac{1}{T} \sum_{t=1}^{T} (y_{t+h} - \hat{y}_{t+h|t})^2}$$
,

with y = normalized DNI and h = 10, 20, 30 min

#### RMSE Table (in W/m2)

	t+10 min	t+20 min	t+30 min
Persistence	160,73	204,40	232,14
ARMA	155,67 (+3,1%)	191,68 (+6,2%)	214,10(+7,8%)
ARMAX	148,62 (+7,5%)	181,61 (+11,11%)	202,76 (+12,7%)



## **Conclusion & Perspectives**

- Total Sky Images help time-series modelling provide **better** solar direct irradiance forecasts.
- Proposed framework can be a starting point to numerous lines of research:
  - Extension of ARMA models: Regime-switching model (TAR) driven by image information.



ARMAX extension: use more than 1 external input (e.g. proportion of thin/thick cloud, kind of cloud, cloud motion...)











#### Thank you!

Any questions?

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