

## A NEURAL NETWORK APPROACH FOR GENERATING SOLAR IRRADIATION ARTIFICIAL SERIES

P.J. ZUFIRIA\*, A. VAZQUEZ-LOPEZ\*, J. RIESCO-PRIETO\*, J. AGUILERA\*\* & L. HONTORIA\*\*

\*Grupo de Redes Neuronales. Dpto. de Matemática Aplicada a las Tecnologías de la Información.

E.T.S. Ingenieros de Telecomunicación. Universidad Politécnica de Madrid.

Ciudad Universitaria s/n. E-28040 Madrid, Spain

e-mail: pzz@gauss.mat.upm.es // jacob@tid.es

\*\* Grupo Jaén de Técnica Aplicada. Dpto. de Electrónica. Universidad de Jaén.

Avda. Madrid, 35. 23071 Jaén, Spain.

e-mail:aguilera@ujaen.es // hontoria@ujaen.es

### OBJECTIVE

GENERATION OF SOLAR IRRADIATION ARTIFICIAL SERIES BY MEANS OF A NEURAL NETWORK APPROACH

#### NATURE OF INFORMATION

DATA → ATMOSPHERIC TRANSMITTANCE OR CLARITY INDEX  
HOURLY VALUES: 9 YEARS x 365 DAYS x 16 HOURS VALUES

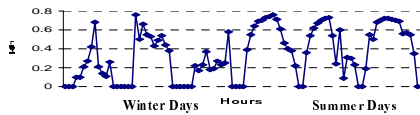
$$k_{th} = G_h / B_{oh}$$

$k_{th}$  = atmospheric transmittance or clarity index during hour  $h$

$B_{oh}$  = extraatmospheric solar irradiation on horizontal surface during hour  $h$

$G_h$  = solar irradiation on the horizontal surface during hour  $h$

Example of Clarity Index evolution in three winter days and in three summer days



#### PROPOSED GENERATION METHOD

\* MAIN FEATURE: ASSOCIATED INFORMATION INCLUDED STEP-BY-STEP

##### STEP 1

SERIES OF  $9 \times 365 \times 16$  VALUES CONSIDERED AS A NUMERIC SEQUENCE

##### STEP 2

INCLUDING DAILY INFORMATION + THE 3 PREVIOUS  $k_{th}$  VALUES

“NORMALIZED DAY  $d_n$ ” DEFINED AS:  $d_n = 1 - (Nd - 163) / 163$

$Nd$  = day number within the year

##### STEP 3

INCLUDING “DAILY CLARITY INDEX  $K_T$ ”: DEFINED AS:  $K_T = G_d / B_{od}$

$B_{od}$  = extraatmospheric solar irradiation on horizontal surface during day  $d$

$G_d$  = solar irradiation on the horizontal surface during day  $d$

##### STEP 4

INCLUDING “HOUR ORDER NUMBER”: DEFINED AS:  $hour_{norm} = (hour - p) / (16 - p)$

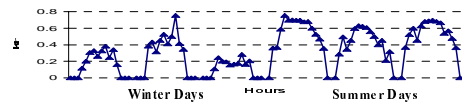
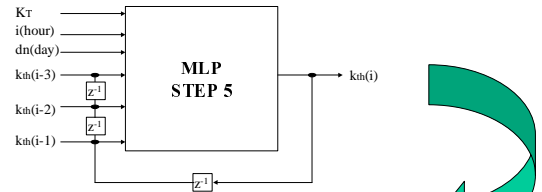
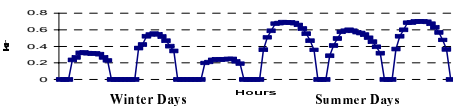
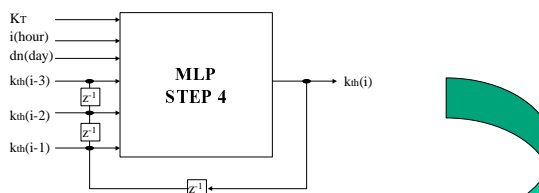
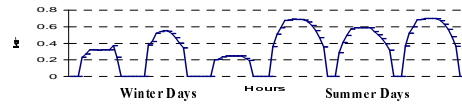
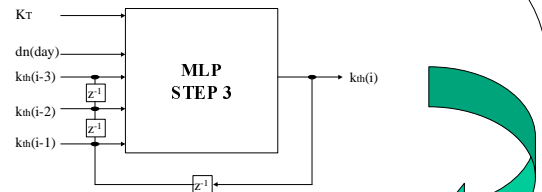
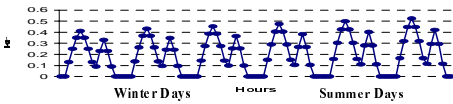
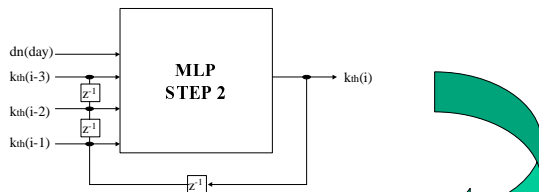
$p$ : prediction window size

##### STEP 5

INCLUDING “STOCHASTIC RIPPLING”.

### EVOLUTION OF THE DIFFERENT RESULTS OBTAINED IN SOME OF THE STEPS

(Each Figure Contains Three Winter Days and Three Summer Days)



### CONCLUSIONS

- \* A methodology based on neural networks has been presented for generating solar irradiation time series.
- \* The generation can be performed with little knowledge of the problem.
- \* The MLP presents the capability for finding relationships among variables with unknown *a priori* relationship.
- \* The proposed method does not assume any *a priori* model as opposed to the standard approximation techniques where polynomial regression techniques are employed.

### FUTURE ACTIONS

- \* Testing the quality of the developed method.
- \* Improvement of the generation of the stochastic rippling component.
- \* Comparisons with other methods of generation solar irradiation time series.
- \* Application of this methodology to other sites where solar irradiation series are available.
- \* Generation of solar irradiation in other time scales as day scale or minute scale.