

Using Adaptive Neuro Fuzzy Inference System to Build Models with Uncertain Data for Rainfed Maize

Study Case in the State of Puebla (Mexico)

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Abstract: Using the methodology of Adaptive Neuro Fuzzy Inference System (ANFIS) a model to determine the relationship suitability index with the yields per hectare and the percentage of production area lost of rainfed maize for the state of Puebla was built. The data used to build the model presented inconsistencies. The data of the INEGI's land use map presented more municipalities without rainfed maize agriculture than the database of SAGARPA. Also the SAGARPA data, in terms of the percentage of production area lost, do not show any distinctions between the loss due to climate, pests, or simply that the farmer did not plant the total area that was declared, or had not harvested all the area declared. Even with data inconsistencies ANFIS produced a coherent output reviewed by experts. The model shows that higher the percentage of production area lost and high yields the higher the suitability index is. According to local studies this is due to the high degradation of the soils.

1 INTRODUCTION

Adaptive Neuro Fuzzy Inference System, or simply ANFIS, is a neural network based on Takagi-Sugeno fuzzy inference system. By integrating both neural networks and fuzzy logic principles, it has the potential to capture the benefits of both in a single framework. It has the ability to construct sets of fuzzy if-then rules to approximate nonlinear functions. ANFIS can also build appropriate membership functions to generate the stipulated input-output pairs to be used in the model. (Jang, 1993) The Neuro-adaptive learning techniques provide a method for building a fuzzy model from the information contained in a data set.

The fuzzy system enables a flexibility in the variables, and representation of incomplete data, as membership to a fuzzy set is denoted by the degree of membership to the set.

Since the ANFIS can deduce relations between the inputs/outputs, ANFIS forms an input output mapping based both on human knowledge (based on fuzzy if then rules) and generated input/output data pairs by using a hybrid algorithm that is the combination of the gradient descent and least square estimates. (Jang, 1993)

In this case study for the state of Puebla, ANFIS was used to determine the relationship between the natural suitability index of rainfed maize, yields and the percentage of production area lost. The index of suitability of rainfed maize was calculated with a fuzzy model base don expert knowledge (Vermonden, 2012) and on the previous work of Monterroso. (Monterroso et al., 2011) The Index had a resolution of 1 km by 1 km. The suitability index is calculated using mean temperatures, mean precipitation, depth of soil and slope. The data of yield per hectare and the percentage of production area lost is presented at municipality level from the Secretaria de Agricultura, Ganadería, Desarrollo Social, Pesca y Alimentación (SAGARPA, 2000-2008).

2 METHOD

The examination of the data obtained from SAGARPA and the INEGI showed inconsistencies. According to the data from SAGARPA only three municipalities in the state of Puebla have no rainfed maize agriculture, Altepexi, Atzala and Zinacatepec,

while the map of land use from INEGI (INEGI, 2005) shows that other 16 municipalities without rainfed agriculture, see Figure 1.

Another assumption made for the land use map for rainfed agriculture is that in all the area presented maize is being cultivated. Since maize is the most important cereal in the Mexican diet and 92% of the farmers in Mexico that own between 0 to 5 hectares, produce 56.4% of the maize, by the rainfed farming practice. Therefore the assumption is the areas presented as rainfed agriculture are rainfed maize agriculture.

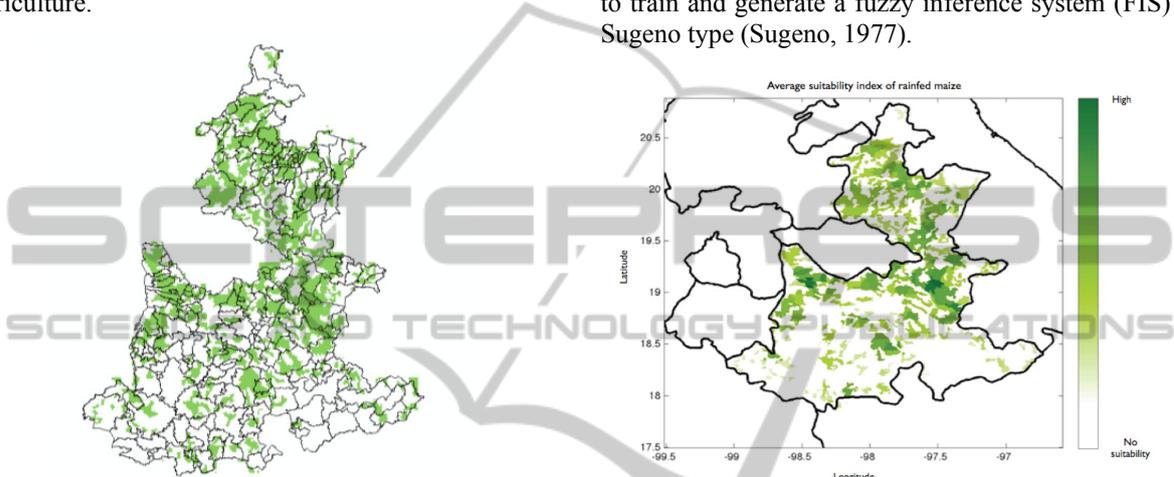


Figure 1: Map of the state of Puebla showing in green the municipalities division and land use for rain-fed agriculture.

Since the data from SAGARPA is at municipality level for the period of 200 to 2008, the map of land uses of INEGI, was used as a mask to extract the data of the suitability index, since it would correspond to the area marked where rainfed maize agriculture is being produced. An average was calculated to obtain the equivalent scale of the following two variables (Figure 2). Figure 3 shows the values of yield per hectare and Figure 4 the percentage of production area lost.

To determine the relationship between the three variables, a subtractive clustering algorithm (Chiu, 1994) was used to generate a fuzzy system. This algorithm allows to estimate the number of clusters and their centers, to latter build the membership functions and the relations between the variables. First the centers are established through subtractive clustering method (Dubois and Prade, 1980), once the center is calculated it determines the radius of influence. For each data of the set a potential measure is calculated to check the center of the cluster using the density of surrounding data. This is done to identify natural groupings of the data from a

large data set, allowing concise representation of relationships embedded in the data. In this case four clusters were calculated, and thus reducing the complexity of the sets and the analysis of the relationship between the variables. The data of the clusters found was used later in the training of the model.

After obtaining the clusters, these were used to generate the if-then rules and membership functions. The information was added to the genfis2 function (MATLAB, 2008) and 75% of the data set was used to train and generate a fuzzy inference system (FIS) Sugeno type (Sugeno, 1977).

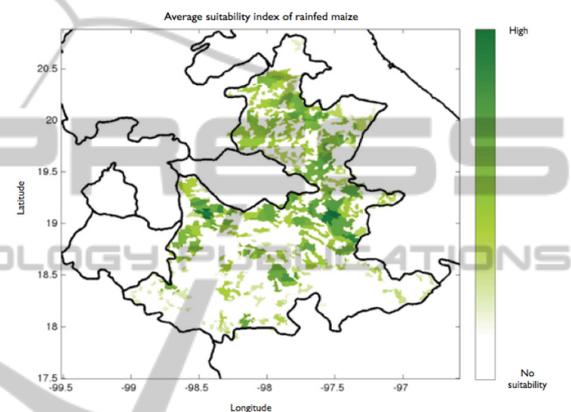


Figure 2: Map of the state of Puebla where the average of the suitability index for rain-fed maize is shown with the mask of land use for rain-fed agriculture

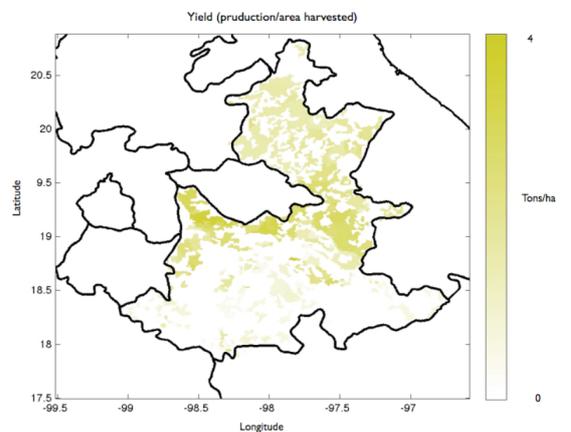


Figure 3: Map of the state of Puebla showing the yield per hectare with the mask of land use for rain-fed agriculture.

First with the information obtained with the subclustering method it determines the number rules and antecedent membership functions and the uses the least square estimation to determine each rule's consequent equations. Returning a FIS structure that

contains a set of fuzzy rules to cover the feature space.

The 25% left of the data set is then used to verify the model. To verify the model the root mean square error of the system generated by the training data was calculated to be 0.0736. The root mean square error of the system was used for both checking and testing the FIS parameters was 0.0947. Both very close to zero.

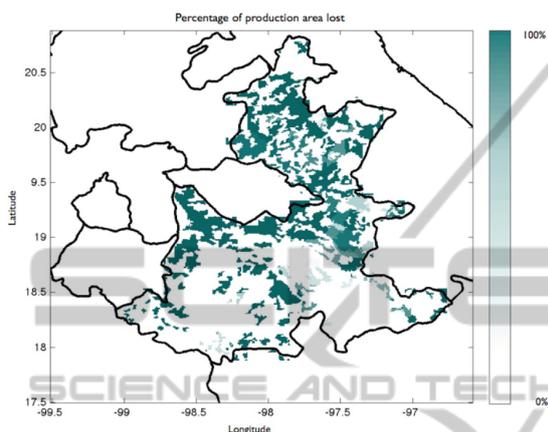


Figure 4: Map of the state of Puebla showing the percentage of production area lost with the mask of land use for rain-fed agriculture.

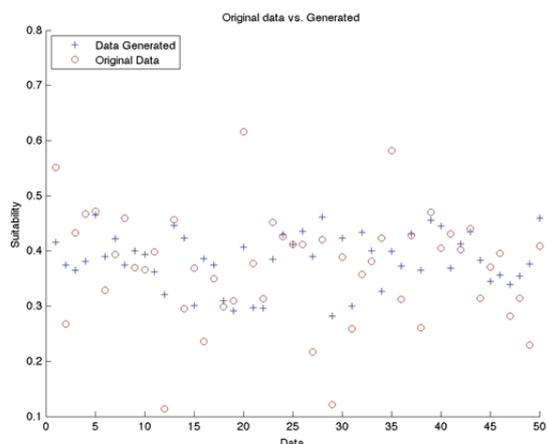


Figure 5: Graph showing the 25% of the data used to verify the model, in circles the original data, the calculated data by the model after improving its capacity with ANFIS in crosses.

ANFIS is used to improve the capacity of the FIS to model the data. Again 75% of the data is used to train the neuro-adaptive network. In this case the hybrid optimization method was used, which combines gradient descent and the least squares method. The gradient descent is used on the premise parameters that define the membership functions and

for the consequent parameters that define the coefficients of each output equations the least squares method is used. A hundred Epochs were used and the training error tolerance was set to zero. Stability of the training was achieved before Epoch 30. To verify the model the root mean square error of the system generated by the training data was calculated to be 0.0710. The root mean square error of the system was used for both checking and testing the FIS parameters was 0.0854. Improving the previous the FIS generated by genfis2.

3 RESULTS

The fuzzy surface of the rules generated with the data (Figure 6) show that the areas with the highest suitability index have the highest percentage of production area lost.

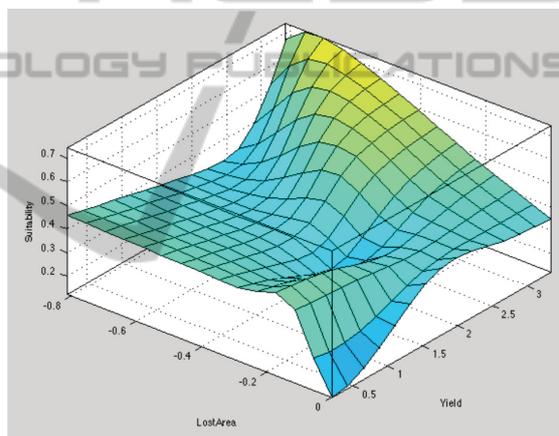


Figure 6: Graph of the surface created by the rules of the FIS.

The FIS generated with the data gave; for the first input variable (percentage of production area lost), four membership functions that have a tendency to the lower losses whilst the membership functions for Yield area better distributed across the interval, but cluster1 and cluster 4 cover a very similar area. As shown in Figure 7.

Four rules were generated:

If (AreaLost is Cluster1) and (Yield is Cluster1) then (Suitability is Cluster1)

If (AreaLost is Cluster2) and (Yield is Cluster2) then (Suitability is Cluster2)

If (AreaLost is Cluster3) and (Yield is Cluster3) then (Suitability is Cluster3)

If (AreaLost is Cluster4) and (Yield is Cluster4) then (Suitability is Cluster4)

This preliminary model can be used to formulate scenarios on how the yield per hectare and the area

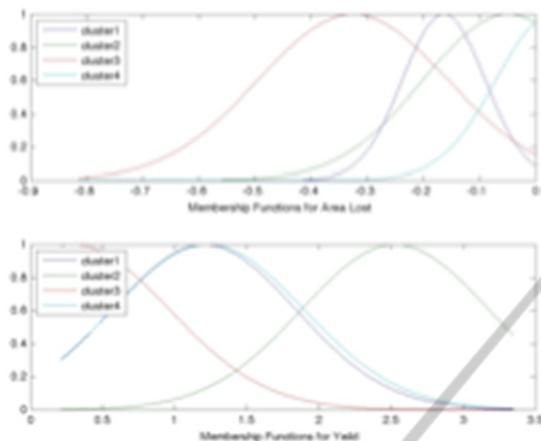


Figure 7: Membership functions generated by ANFIS.

of production loss due to the change in the suitability index by climate change since two of the variables used to build it are mean temperature and precipitation.

The results of the model show an important human hidden factor in the data, since a farmer can declare production areas lost to claim insurance or simply didn't plant the area he declared, which is reflected on the surface of figure 6, as well as in the membership functions for this variable as they are all in the same range, where high percentage of the production area is lost and medium yield production should have a low suitability index.

4 CONCLUSIONS

The state of Puebla is known for the origin of cultivated maize. The methodology used was the subtractive clustering analysis and ANFIS to establish the relationships between the suitability index for rain-fed maize and the other variables. This preliminary model reflects where suitability is higher then the area lost is higher. A study of the municipality of Molcaxac (Gaspar Angeles et al., 2010), which has a high suitability index for the period of 2002 to 2003 only cultivated 35% of the total production of the cereal, due to the degradation of the soils. The data of SAGARPA has a few inconveniences since they are presented at the municipality level and within the same municipality the range in suitability index may present high variations. Also the SAGARPA data, in terms of percentage of production area loss, do not show any

distinctions if the loss was due to climate, pests, or simply that the farmer did not plant the total area that had been declared, or hasn't harvested all the area declared (which can occur when the price of corn falls and no longer compensates the harvesting cost). The data obtained is from 2000 to 2008, since in older data the number of municipalities decreased (since new municipalities are created) and much older data is only at the rural development districts (DDR) level, which do not have a clear idea of the municipalities belonging to each one, and some may even belong to several, nor there is a map of them adding more uncertainty to the model.

This model shows that agriculture as any human system is complex, and it requires a greater number of variables in order to make the results more understandable. These variables could be the use of fertilizer, pesticides, enhanced maize seeds, soil degradation. Also interviews with farmer could ameliorate the results and determining which areas on the map are being used for maize and which are not, this would also help understand why the high suitability areas have the highest losses. But preliminary results allow us to establish relationships between these variables that experts find coherent and that more detailed studies like the study of the Molcaxac municipality are showing to be an alarming trend in the state of Puebla.

This kind of model can simplify the decision making process since the results are objective and transparent based in mathematical principles, and the results of this model are significant even if the data is insufficient, helping to understand reality better.

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