

REVIEW 1

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- Milos Ilic, Sinisa Ilic, Srdjan Jovic, Stefan Panic,
- Early cherry fruit pathogen disease detection based on data mining prediction", 2018, Computers and Electronics in Agriculture (Elsevier).
- Impact Factor **2.427**

- Introduction
- Problem Statement
- System Model
- Data Visualization
- Program for Image Prepossessing
- Program for Edge Detector
- Images of Edge Detector
- Result Discussion
- Conclusion
- Future Work
- References

- First goal of this paper is **Data prepossessing, Visualization, Outlier detection** and Reduction of redundancies among instances.
- Second goal is to create prediction models based on refined data after Prepossessing by using different **data mining prediction techniques**.
- Third goal is comparison of **prediction models** and **selection of the best model**.
- Determined by the basis of **statistical errors** and **data mining errors** for the each model
- Focuses on two data sets such as, First data set type is containing data used for **prediction model training**
- Second data set type represents data that are used for **Prediction model testing** and **Accuracy evaluation**

- Different Mathematical techniques for **Data Processing** and **Data Protection**
- Important diseases in cherry Fruit is **Monilinia Laxa** and **Coccoomyces Hiemalis**
- Achieved Prediction accuracy is **95.8%**

- Prediction can be performed based on weather data and data that represent **specific plant or disease** characteristic that are maintained
- Weather conditions are favorable for disease development which can greatly reduce the problems and helps in **fruit protection**
- By the **least expensive** for collection are weather parameters and each disease can **infect fruit tree** if specific weather conditions are satisfied
- Weather factors are selected in **every month** as prediction parameters.

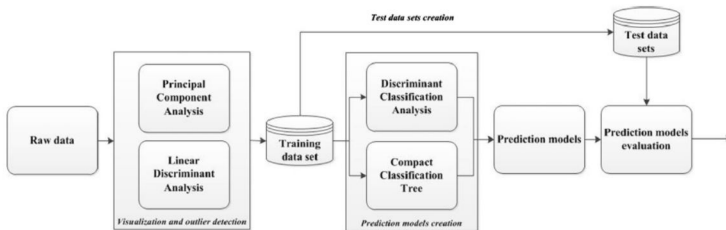


Figure: System Model for Mutual relations for implemented Method

Source: Computers and Electronics in Agriculture (Elsevier)-2018

- Data prepossessing
- Evaluation prediction model creation
- Prediction of future class attribute values

- File System used here is **single .xlsx file** ,which provides input to the System Visualization and Outlier Detection module
- Two groups of methods **PCA** and **Discriminant analysis**.
- Some **outliers** are detected using PCA and remaining using **Discriminant analysis**.
- Each of them has own advantages and flaws

- Varies types of analysis are present in both **Classification and Clustering**
- Deals with tree Structure (i.e) Classification tree, Decision tree
- Developed predictive models are mainly evaluated classification processing.
- Accuracy of prediction is evaluated by **Comparing the actual value of the class attribute** and the value of predicted class attributes in the test data set.

- It mainly used abundantly in various analysis such as **Neuroscience to Computer graphics** because it is a simple, non-parametric method of extracting relevant information from raw data sets.
- Additional effort PCA provides a roadmap to **reduce a complex data set** to a lower dimension to reveal the sometimes **hidden, simplified structure format**.
- Significance of the **DPCA matrix** is having high dimensional data and reduce it to just a few principal components hopefully.
- Capture variations of the data and allow **Visualization, Patterns and Structures**

- Discriminant analysis consist of a **instances of observations** with known **group membership** together with their continuous variables
- Prior probability can have one of the three choices: **Uniform, Empirical** and **Numeric vector**.
- Uniform prior probability of class k is 1 over the **total number of classes** present.
- Empirical prior probability of class k is the number of training instances of **class k** divided by the **total number of training instances**.
- Numeric vector prior probability of class k is the **j th element of the prior vector**.

- Classification tree splits nodes either impurity or node error, but only three of them are widely used **Gini splitting rule, Deviance** and **Twoing splitting rule**
- Decision trees are then used to **classify new data**. They are used for each instance from learning data set the class value is known.
- It may be provided by user or calculated in accordance with **some exogenous rule**
- Common Notations are used as, **tp** be a parent node and **tl, tr** for left and right child nodes of parent node(tp).
- Variable matrix **X** with **M** number of variables x_j and **N** observations made in it.

Variable name	Min	Max	Mean	Variance	StdDev
Min temp	-2	28	12.97	24.52	4.95
Max temp	5	26.26	47.14	6.87	40
Mean temp	3	33	19.60	29.93	5.47
Average humidity	0.33	0.97	0.68	0.02	0.13
Rainfall	0	30	1.05	10.56	3.25
Wind speed	3	27	7.59	16.63	4.08

Tabulation:Evaluation Parameter

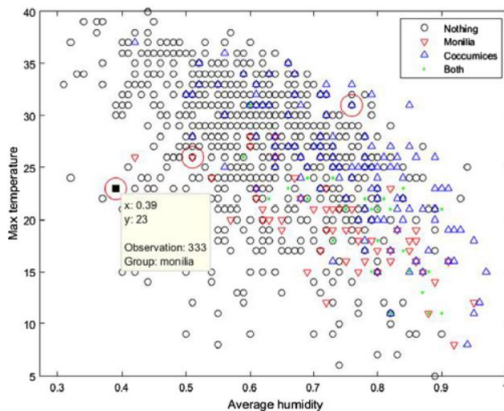


Figure: 2D Visualization of Training Data Set

Source: Computers and Electronics in Agriculture(Elsevier)-2018

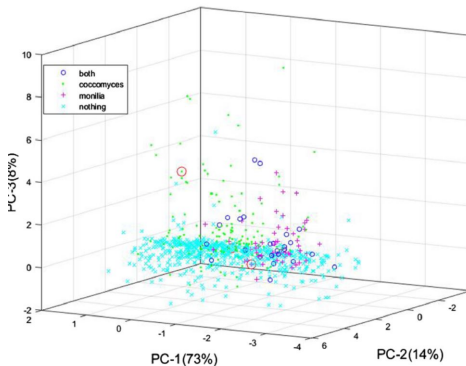


Figure: 3D Visualization

Source: Computers and Electronics in Agriculture(Elsevier)-2018

Program for Image Preprocessing

```
from PIL import Image
import PIL
from PIL import Image

image=Image.open("C:\\Users\\Aravind Prasad\\Desktop\\MRICE\\Bacterial leaf blight\\a.jpg")
print(image.format)
print(image.mode)
print(image.size)
image.show()
image.thumbnail((100,100))
print(image.size)
cropped=image.crop((100,100,300,200))
cropped.show()
cropped.show()
cropped.show()
```

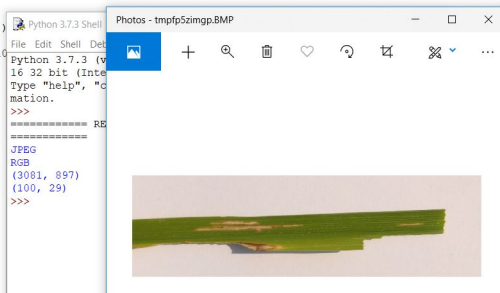


Figure: Image Preprocessing

Program for Edge Detector

```
import cv2
from matplotlib import pyplot as plt
img = cv2.imread('C:\\Users\\Aravind Prasad\\Desktop\\MRICE\\Bacterial leaf blight\\d.JPG')
grey = cv2.cvtColor(img,cv2.COLOR_BGR2GRAY)
blur = cv2.GaussianBlur(grey, (5, 5), 0)
edge = cv2.Canny(blur,100,200)
dil = cv2.dilate(edge, None)
plt.subplot(231),plt.imshow(img,cmap = 'gray')
plt.title('Original Image'), plt.xticks([], plt.yticks([]))
plt.subplot(232),plt.imshow(grey,cmap = 'gray')
plt.title('Grey Image'), plt.xticks([], plt.yticks([]))
plt.subplot(233),plt.imshow(blur,cmap = 'gray')
plt.title('Blur Image'), plt.xticks([], plt.yticks([]))
plt.subplot(234),plt.imshow(edge,cmap = 'gray')
plt.title('Edge Image'), plt.xticks([], plt.yticks([]))
plt.subplot(235),plt.imshow(dil,cmap = 'gray')
plt.title('Dilate Image'), plt.xticks([], plt.yticks([]))
plt.show()
```

Figure: Program for Edge Detector

Images of Edge Detector

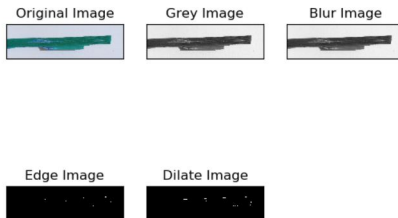


Figure: Edge Detector

- Re substitution error is error rate obtained as **0.0286** and for 10-fold cross validation is **0.0817**.
- Trees that classifies the original training set well, but the structure of the tree is sensitive and its performance on new data is likely to **degrade** for **test data set**.
- Easier for **finding simple tree** for unknown data set performs better than a complex one.

- Predicting time for infection occurrences, the chemical protection is **more efficient** so that **farmers could save money** and on the other side more importantly, it **provides healthier food** due to reduced chemical treatments.
- Implemented algorithms are used for **infection prediction** on the two specific cherry fruit diseases.
- Evaluation of all implemented algorithms was performed under the **same conditions** and the **same evaluation parameters** were observed.
- Main parameter for the development of the disease is **minimal rainfall** (i.e) if there is no rain, necessary conditions for the development of the disease are not fulfilled.

- Authors future plan is to extend the tool with **automated weather data collection** through the network of meteorological stations.
- Another upgrade in the future will be **automatic farmer notification of possible infection** through the mobile or Internet network.

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THANK YOU