

# Work Summary

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Thursday 25 July 2024



# Introduction

- 1 Explanation Of Dataset
- 2 Indication Of Infected Thermogram
- 3 Feature Extraction
- 4 Classification
- 5 Summary and Inferences
- 6 References

## Problem Statement

Classification of Diabetic Foot Ulcer Using Thermograms

# Feature Extraction

- 1 Textural Features
  - GLCM Properties
- 2 Histogram Features

# GLCM

## Defination

- It is a matrix
- It stand for
- **Gray-Level Co-occurrence Matrix**
- Here **Gray-level** refers values of gray intensity in Source grayscale image.
- **Co-occurrence Matrix** refers to each value in matrix is a frequency where any gray-level value occurs in pair.
- Here a Pair is Determined by two parameters  $d$  (Distance),  $\theta$  (angle)

## Purpose

- It represent probability distribution of occuring pair in image
- Using this Various Properties can be calculated that gives some insight into textural variation of the image
- Properties are
  - 1 Contrast
  - 2 Homogeneity
  - 3 Entropy
  - 4 Correlation
  - 5 Energy/Angular Second Moment

## GLCM Properties: Contrast

### Formula

$$Contrast = \sum_{i,j} (i - j)^2 GLCM(i, j)$$

### Summary

Contrast  $\propto$  *LocalVariation*

### Significance

- Captures local variation in neighbouring pixels.
- **High Contrast:** glcm's values have significant intensity change in neighbour pixel, i.e sharp edges.
- **Low Contrast:** Uniformity in Texture with minimal change in intensity.
- also known as inertia/variance

# GLCM Properties: Contrast



Figure 1: Image with Low Contrast



Figure 2: Image With High Contrast

## GLCM Properties: Homogeneity

### Formula

$$Homogeneity = \sum_{i=0}^{L-1} \sum_{j=0}^{L-1} \frac{P(i, j)}{1 + |i - j|}$$

### Significance

- measures how close the distribution is to the GLCM Diagonal
- also known as Inverse Difference Moment

## GLCM Properties: Correlation



## GLCM Properties: Energy

# Histogram Features

## Mean

$$\mu = \sum_{g=0}^{L-1} gP(g)$$

## Entropy

$$Entropy = - \sum_{g=0}^{L-1} P(g) \log P(g)$$

## Variance

$$\sigma = \sum_{g=0}^{L-1} (g - \mu)^2 P(g)$$

## Skew

$$Skew = \frac{1}{\sigma^3} \sum_{g=0}^{L-1} (g - \mu)^3 P(g)$$

# Histogram Features

## Kurtosis

$$\gamma_2 = \frac{\sum_g (g - \mu)^4 \cdot P(g)}{\left(\sum_g (g - \mu)^2 \cdot P(g)\right)^2} - 3$$

# Classification

- All described features used to classify between a Healthy and an Infected Foot.
- Classifiers were used
  - ① Support Vector Machine.
  - ② K Nearest Neighbour
  - ③ Random Forest
  - ④ Decision Tree

## Results: Accuracy

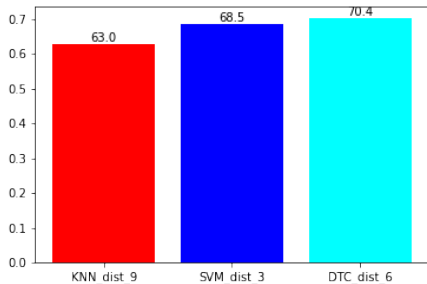


Figure 3: Accuracy

### Description

SVM	KNN	DT
63%	68.5%	70.4%

# Results: Precision

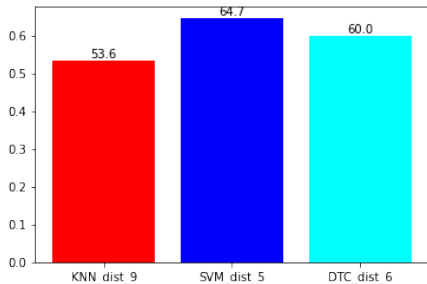


Figure 4: Precision

## Description

SVM	KNN	DT
53.6%	64.7%	60.0%

## Results: Recall

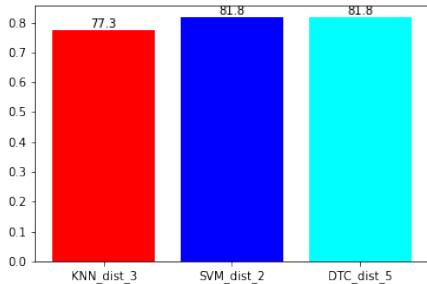


Figure 5: Recall

### Description

SVM	KNN	DT
77%	81.8%	81.8%

## Results: F1-Score

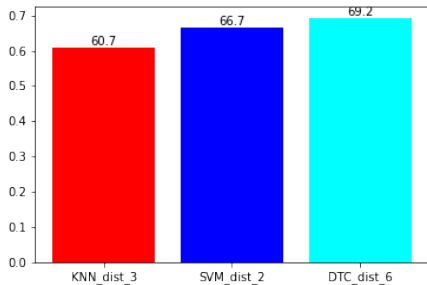


Figure 6: F1-Score

### Description

SVM	KNN	DT
60.7%	66.7%	69.2%



## Results: Accuracy

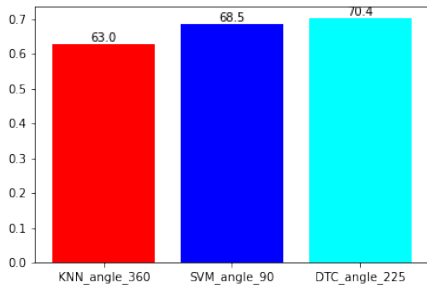


Figure 7: Accuracy

### Description

SVM	KNN	DT
63%	68.5%	70.4%

# Results: Precision

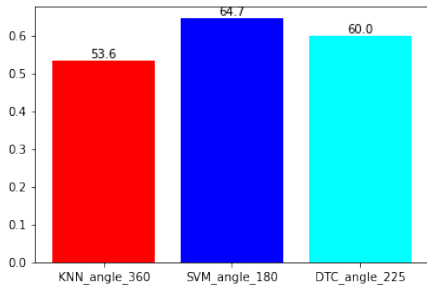


Figure 8: Precision

## Description

SVM	KNN	DT
53.6%	64.7%	60.0%

## Results: Recall

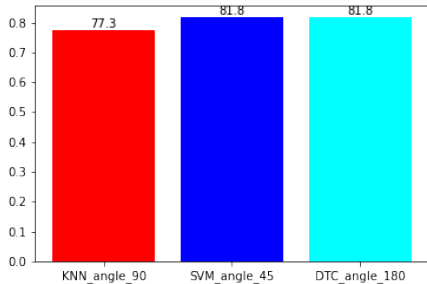


Figure 9: Recall

### Description

SVM	KNN	DT
77.3%	81.8%	81.8%

## Results: F1-Score

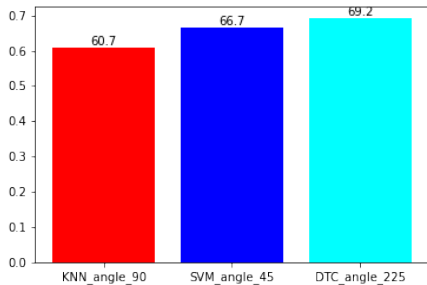


Figure 10: F1-Score

### Description

SVM	KNN	DT
60.7%	66.7%	69.2%

# Results: Variations (Distances)

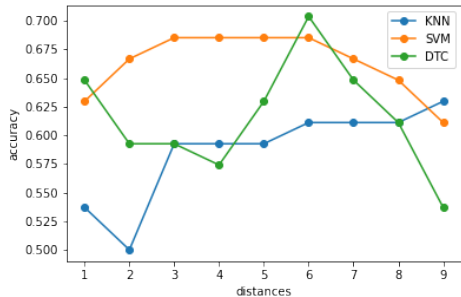


Figure 11: Accuracy Variation

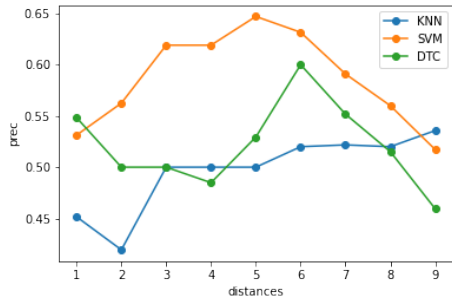


Figure 12: Precision Variation

## Results: Variations (Distances)

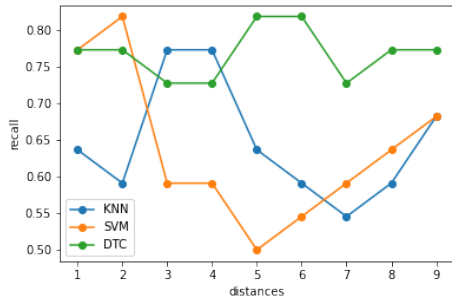


Figure 13: Recall Variation

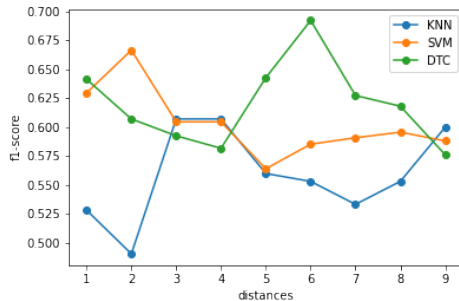


Figure 14: F1-Score Variation

## Results: Variations (Angle)

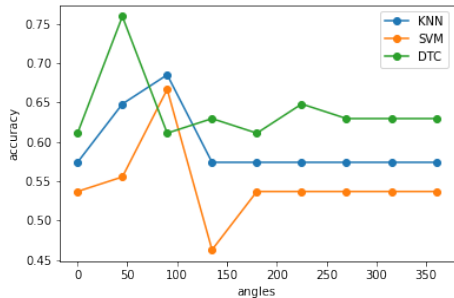


Figure 15: Accuracy Variation

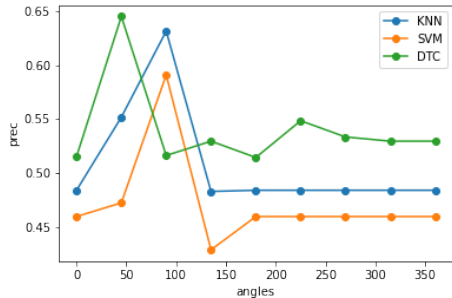


Figure 16: Precision Variation

## Results: Variations (Angle)

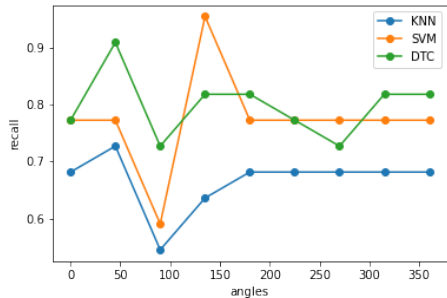


Figure 17: Recall Variation

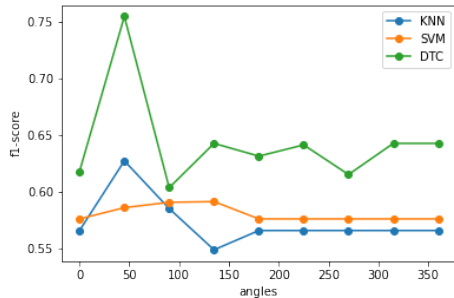


Figure 18: F1-Score Variation



# Conclusion

- From above results it can be seen that maximum acquired accuracy is 70% by **Decision Tree Classifier**

# References

- 1 Eid, Marwa M., Reem N. Yousef, and Mohamed A. Mohamed. "A proposed automated system to classify diabetic foot from thermography." Int. J. Sci. Eng. Res 9 (2018): 371-381.