



Advancing **Diabetic Retinopathy** Detection for Enhanced Eye-Health

Pluridisciplinary Faculty of Nador - SDSI

Project proposed by our Pr. Anas EL ANSARI

Presented by

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Diabetic Retinopathy Detection Introduction

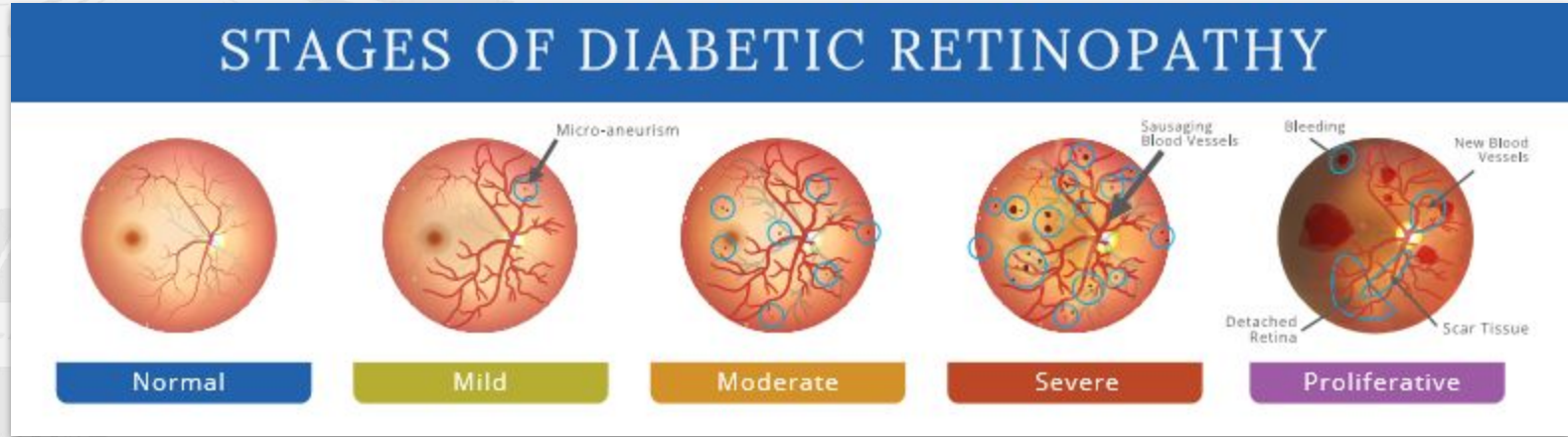
By Issam SEDDIK

Diabetic Retinopathy Detection Introduction

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Problem Illustration

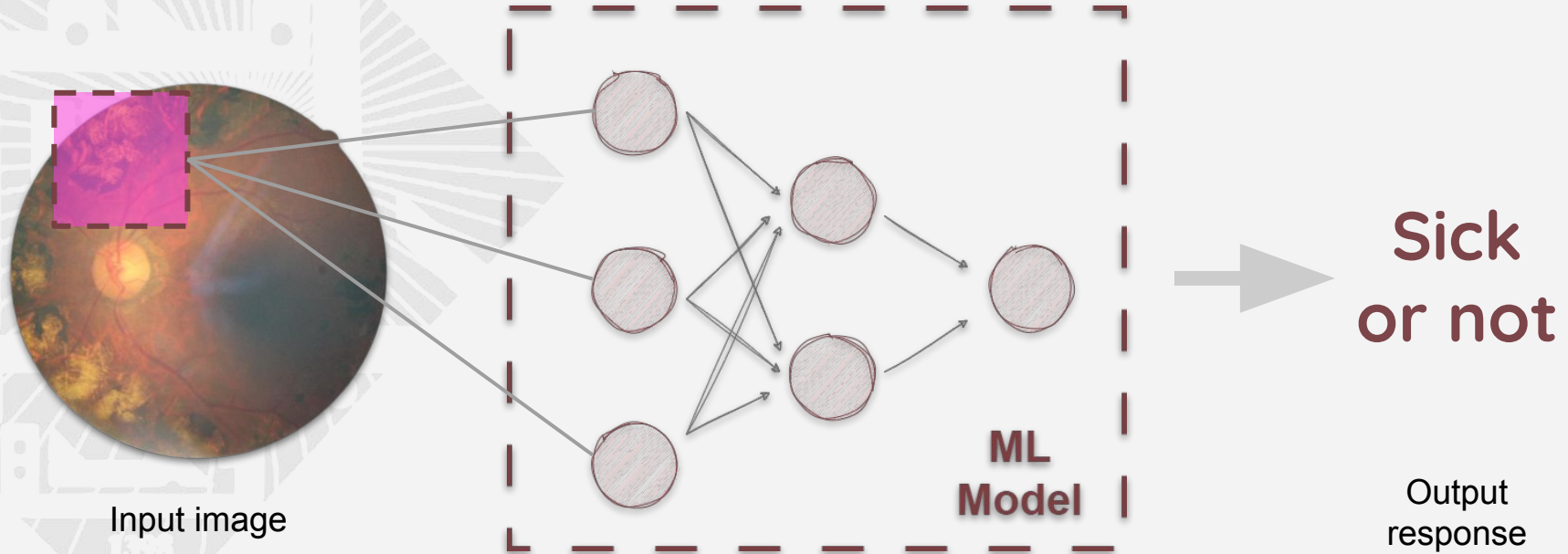
- **Diabetic retinopathy** is a serious eye condition that arises as a complication of **diabetes** mellitus and remains a significant cause of vision impairment and blindness worldwide.
- It affects the **blood vessels in the retina**, the light-sensitive tissue at the back of the eye responsible for vision.



Diabetic Retinopathy Detection Introduction

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Goal



Diabetic Retinopathy Detection Introduction


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 Dataset Card

- The dataset contains radio images of both right and left eyes, sorted into five classes based on the severity of diabetic retinopathy (DR).
- The classes are represented as follow: [no signs of DR](#), [Mild](#), [Moderate](#), [Severe](#), [Proliferative DR](#)

Dataset Id in kaggle [amanneo/diabetic-retinopathy-resized-arranged](#)

Dataset:



No signs of DR	Mild	Moderate	Severe	Proliferative DR
25810	2443	5292	873	708

2.



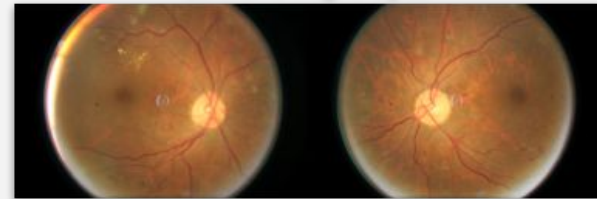
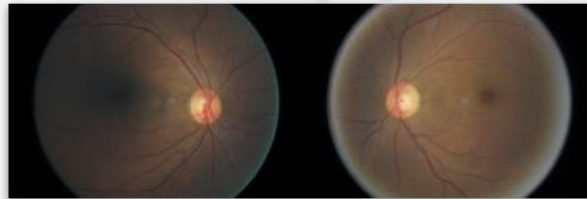
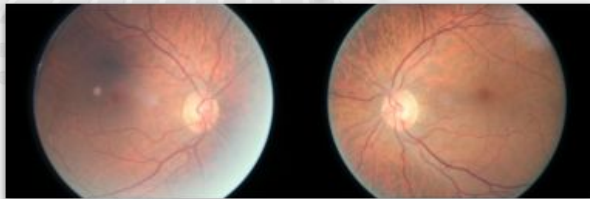
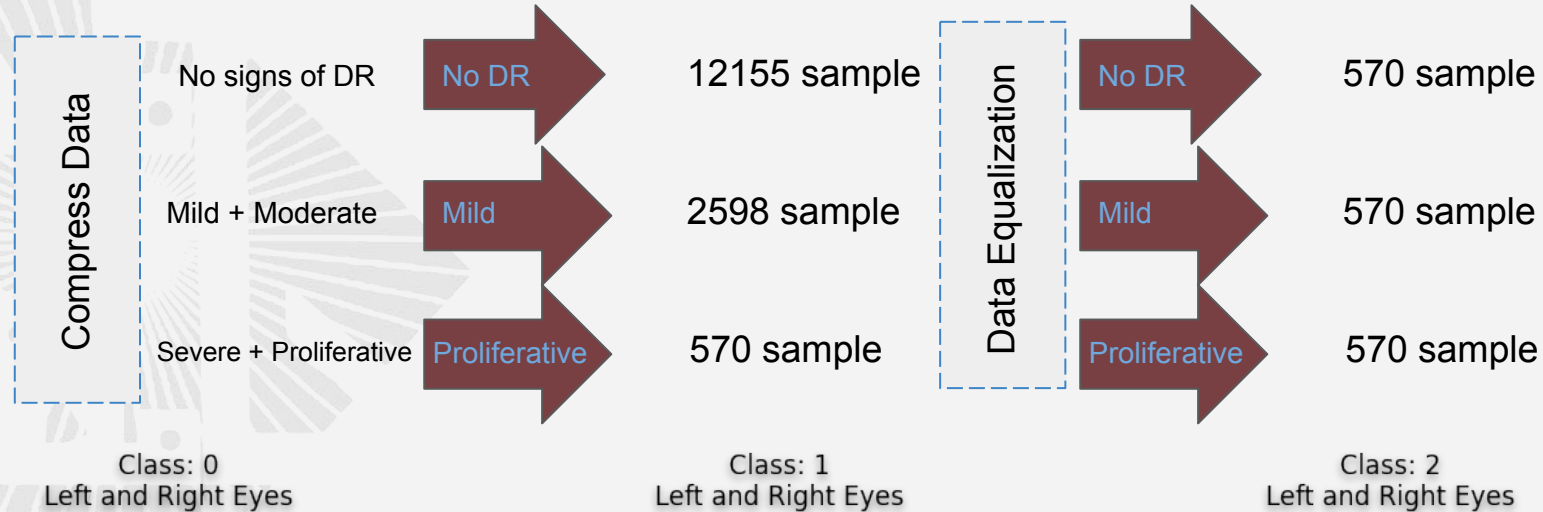
Data Understanding, Compression, and Equalization

By Issam SEDDIK

Data Understanding, Compression, and Equalization

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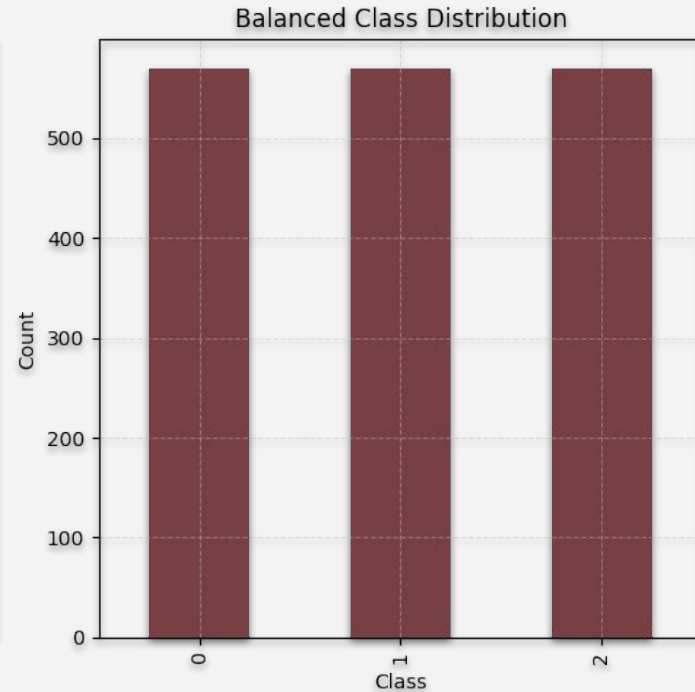
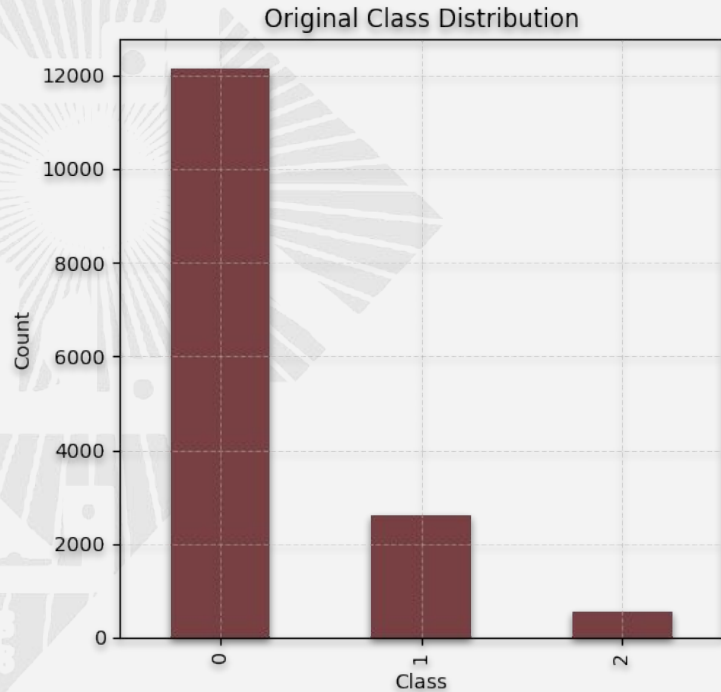
 Data Compression



Data Understanding, Compression, and Equalization

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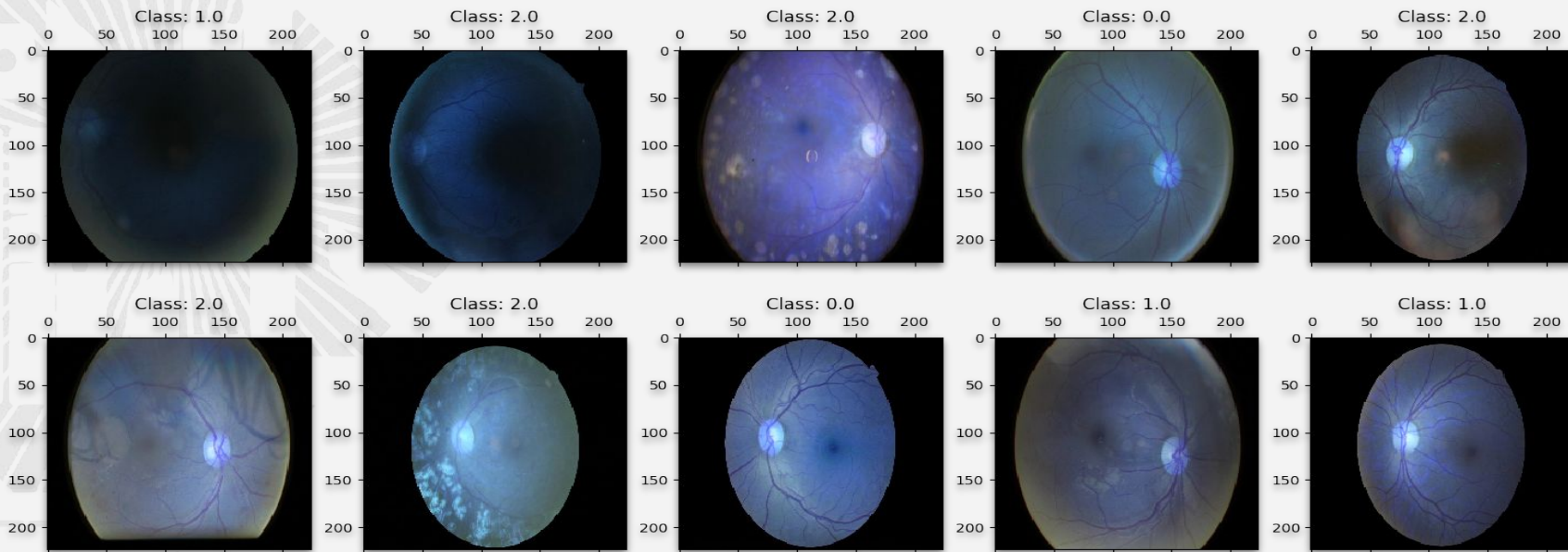
  Data Equalization



Data Understanding, Compression, and Equalization

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Random Sampling

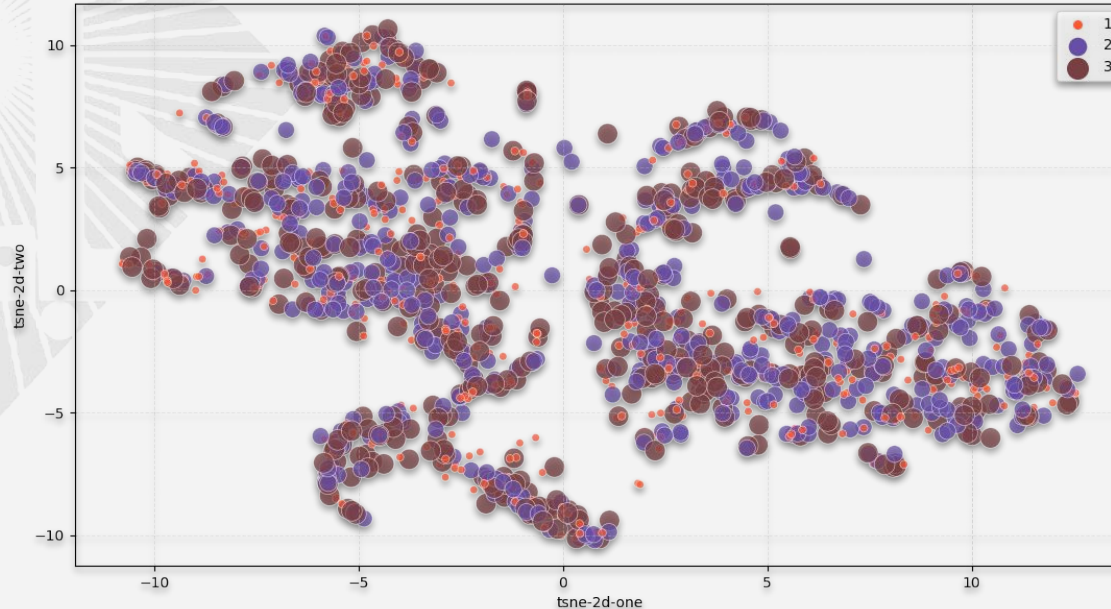


Data Understanding, Exploration, and Pre-processing

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 Data Pre-processing

→ Data visualisation using the T-SNE algorithm



3.



Model Evolution and Selection

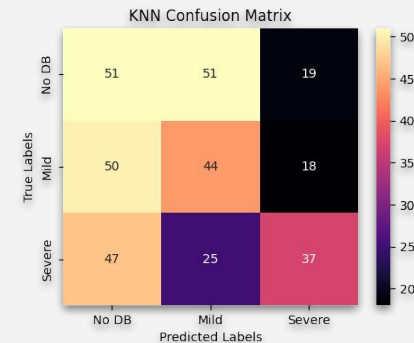
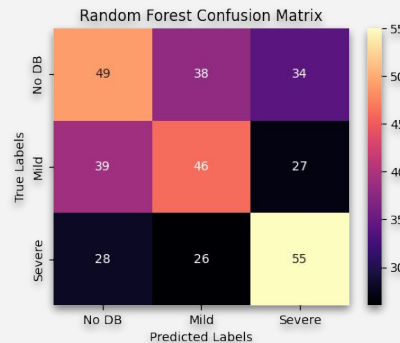
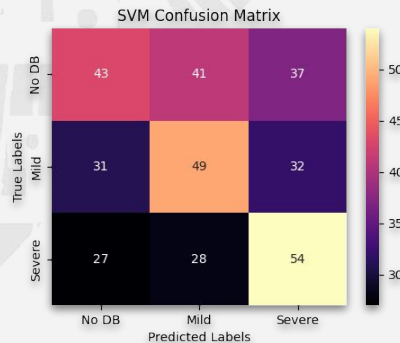
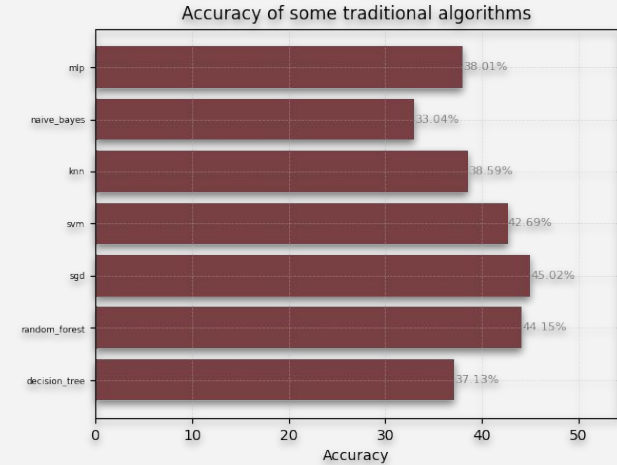
By Jamal BOUSSOUF

Model Evolution and Selection

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Traditional ML Algorithms

→ We have used various traditional ML algorithms to predict the case and these algorithms are not able to predict the case.

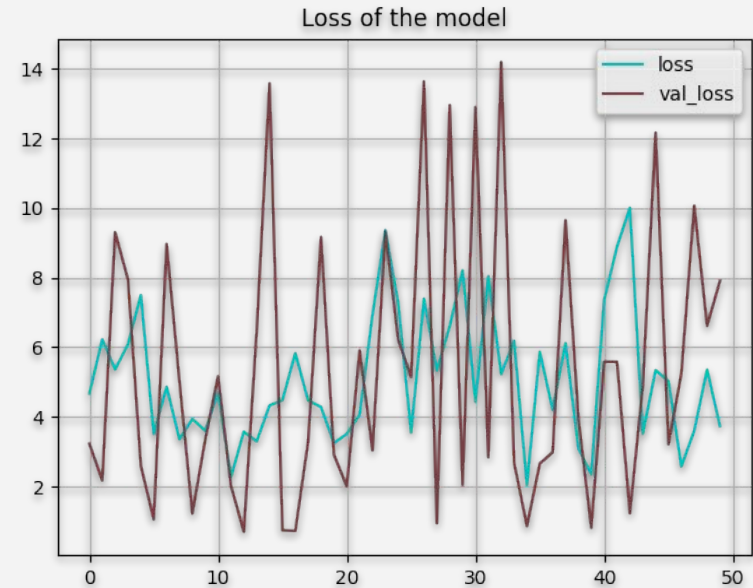


Model Evolution and Selection

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ResNet Model Solution

→ We've also used a large-scale architecture, the ResNet Image architecture, but this doesn't deal with the problem as well as it could!

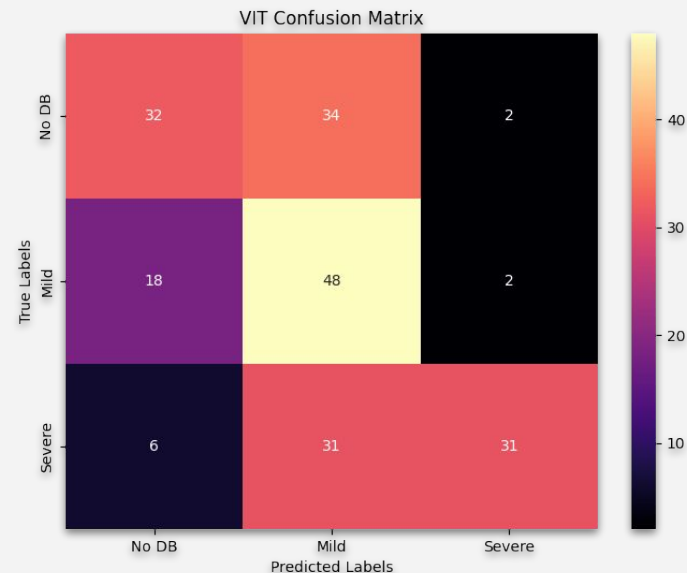
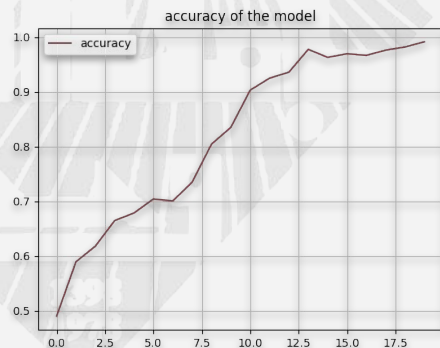


Model Evolution and Selection

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VIT Model Solution

→ Finally, we proceeded with the idea of ViT, which is the type of transformer technique that allows us to achieve better performance through the training phase!



98% / Training set
61% / Testing set

4.



Deployment and set-up

By Jamal BOUSSOUF

API Integration & Prompting with Large Models

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 API Integration

DIABETIC RETINOPATHY PREDICTION

Drag and drop or click to upload the left image

Drag and drop or click to upload the right image

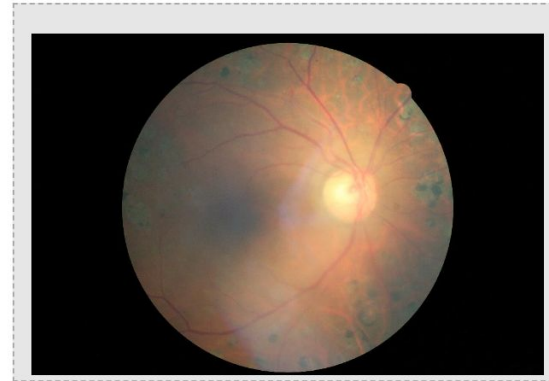
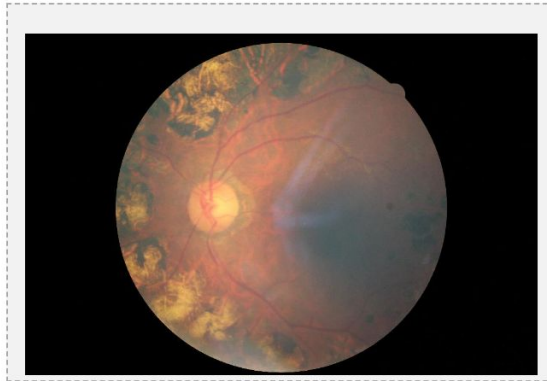
ClearPredict

API Integration & Prompting with Large Models

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 Select LLM

DIABETIC RETINOPATHY PREDICTION



Clear

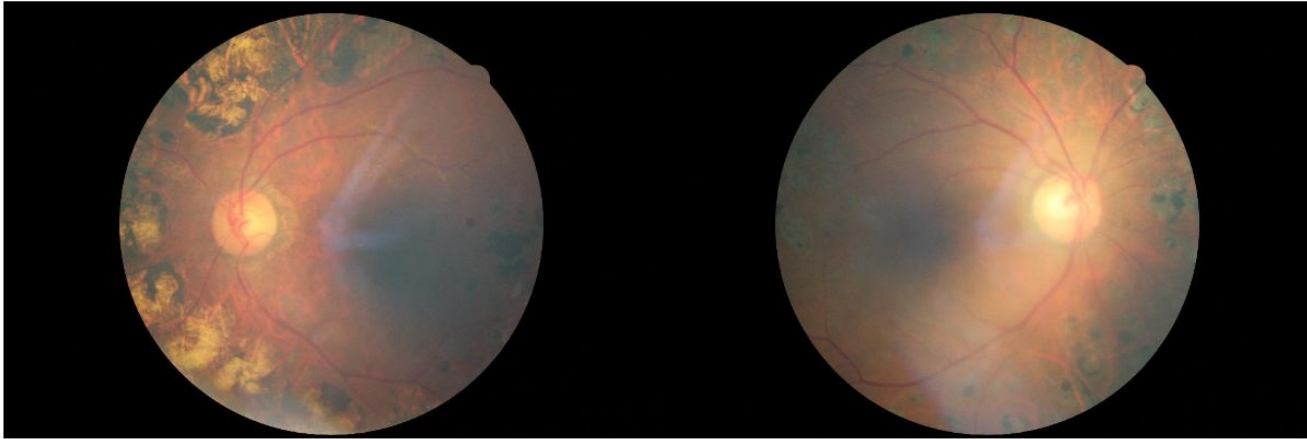
Predict

Chatbot Interface and Interactive Example

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 Chatbot interface

DIABETIC RETINOPATHY PREDICTION - RESULT



Prediction:

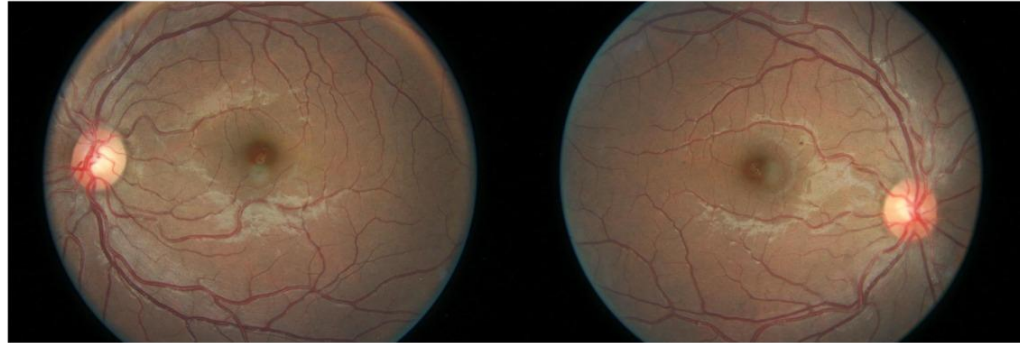
The eyes have been classified as: Proliferative

Chatbot Interface and Interactive Example

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 Chatbot interface

DIABETIC RETINOPATHY PREDICTION - RESULT



Prediction:

The eyes have been classified as: Mild

We hope that you will see a specialized ophthalmologist due to the inability of the model to extract information



Interactive Illustration

5.



Conclusion

By Issam SEDDIK

Conclusion

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Data Understanding

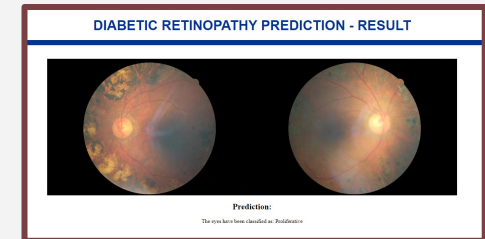
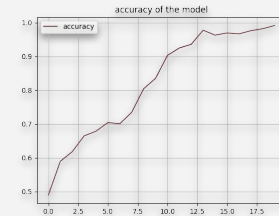
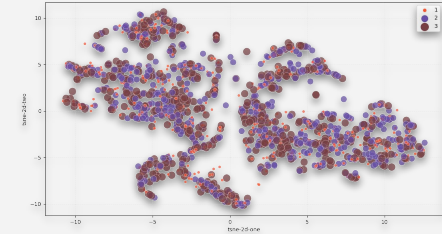
- Understand the data as well as possible and prepare it to be a perfect input of our models.

Modeling

- Find the better model that provide good performance.

Results

- Implement the desired **application**.



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- [1] **Yuan**, Li, et al. "**Tokens-to-token vit: Training vision transformers from scratch on imagenet.**" Proceedings of the IEEE/CVF international conference on computer vision. 2021.
 - [2] **Webb**, Geoffrey I., Eamonn Keogh, and Risto Miikkulainen. "**Naïve Bayes.**" Encyclopedia of machine learning 15.1 (2010): 713-714.
 - [3] **Bala**, Rajni, and Dharmender Kumar. "**Classification using ANN: A review.**" Int. J. Comput. Intell. Res 13.7 (2017): 1811-1820.
 - [4] **Rigatti**, Steven J. "**Random forest.**" Journal of Insurance Medicine 47.1 (2017): 31-39.
 - [5] **Targ**, Sasha, Diogo Almeida, and Kevin Lyman. "**Resnet in resnet: Generalizing residual architectures.**" arXiv preprint arXiv:1603.08029 (2016).



Thank you!

FPN - SDSI - Data Mining

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