

Universität Stuttgart
Institut für Signalverarbeitung und Systemtheorie
Prof. Dr.-Ing. B. Yang

Project I

Diabetic Retinopathy Detection

Deep Learning Lab
Winter-term 2020/2021

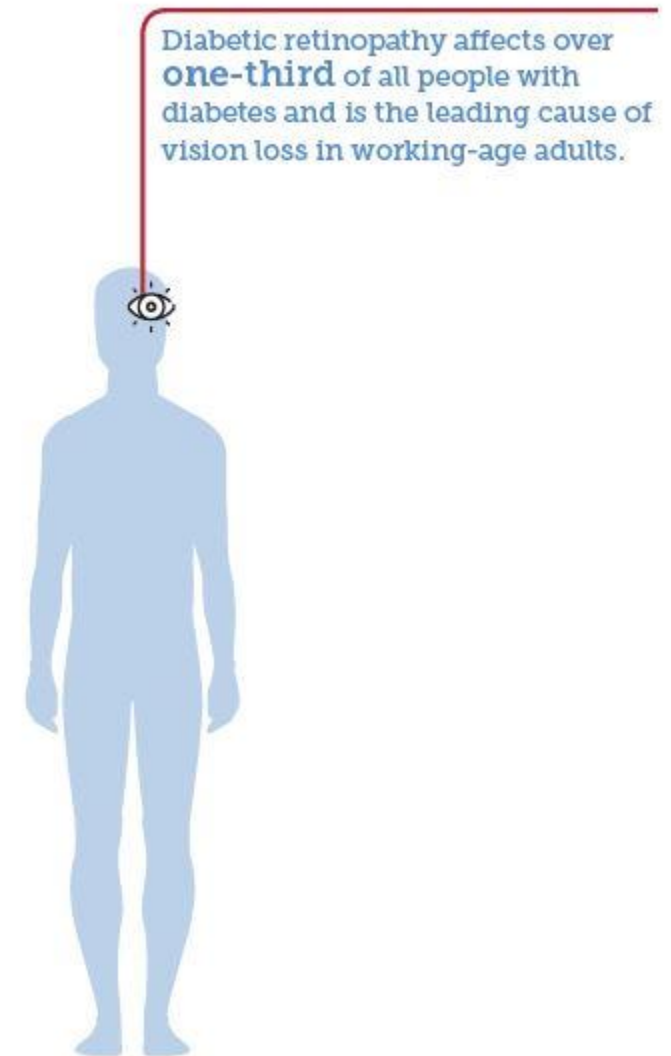
Team

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Swetha Lakshmana Murthy (st169481)

What is Diabetic Retinopathy (DR)?

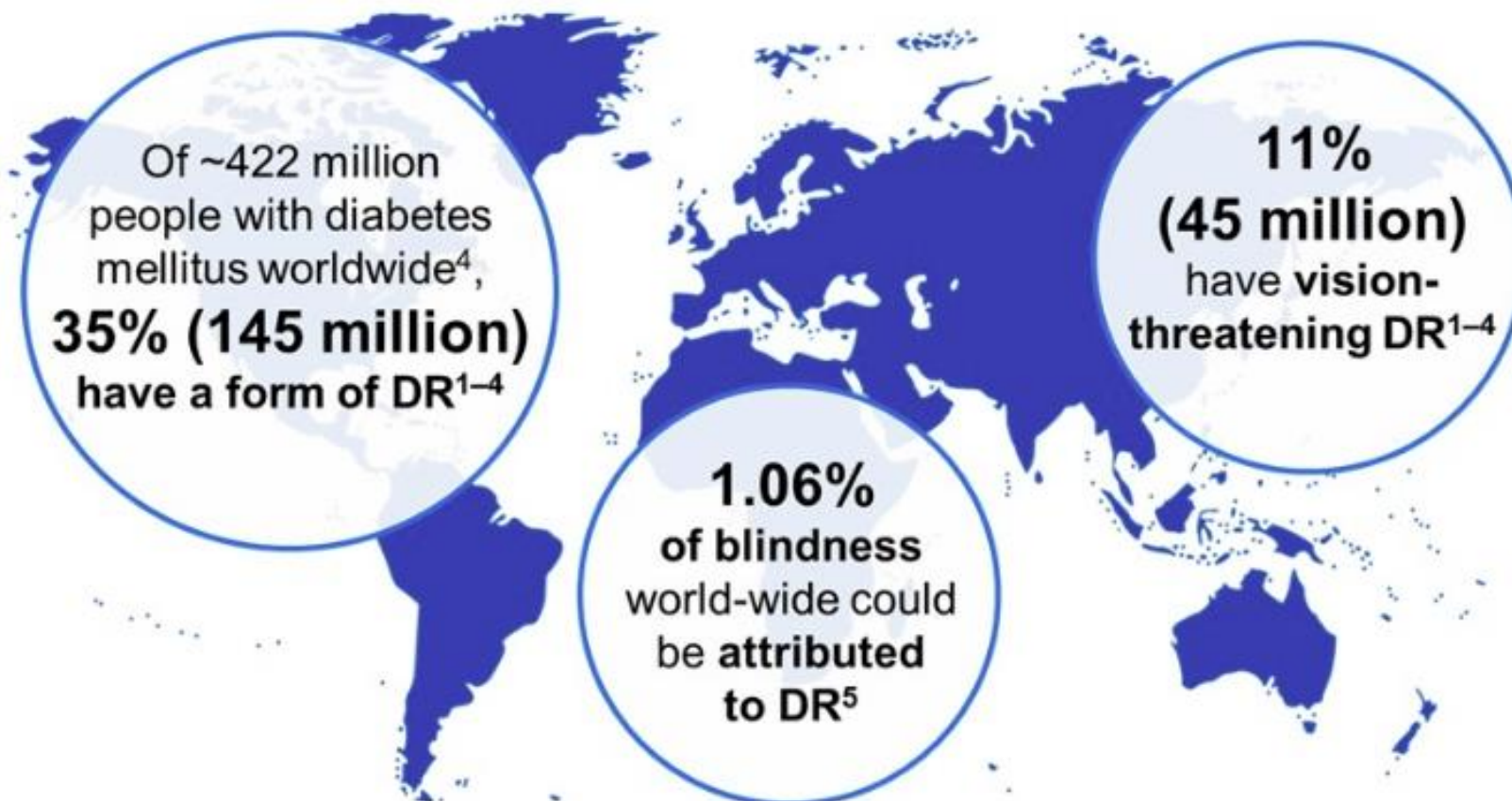
- Abnormal growth of blood vessels in the retina
- Impediment caused to the back of the eye
- Complications resulting in vision impairments
- Results in vision loss if not detected in the early stage



* Source <https://idf.org/our-activities/care-prevention/eye-health.html>

* Please note that Diabetic Retinopathy will be addressed as DR in the presentation henceforth.

Diabetic retinopathy (DR) is the primary cause of vision loss in adults aged 20–74 years^{1–3}



...these numbers are estimated to rise, as the prevalence of diabetes increases^{1,3–5}

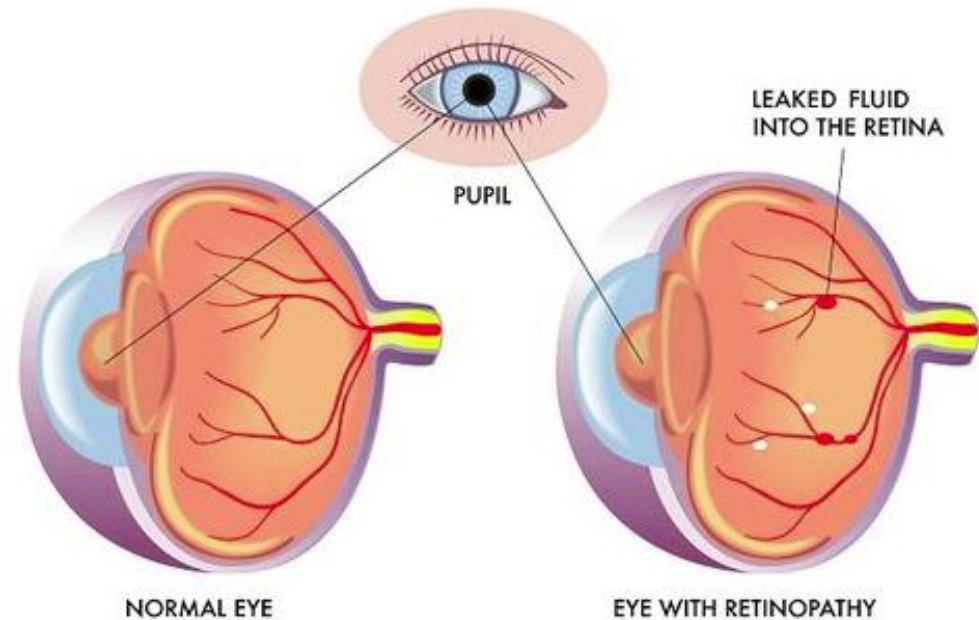
DR, diabetic retinopathy

1. Lee R, et al. Eye Vis (Lond). 2015;2:17; 2. Ting DSW, et al. Clin Exp Ophthalmol. 2016;44:260–77; 3. Yau JWY, et al. Diabetes Care. 2012;35:556–564;
4. IAPB Vision Atlas. Diabetic Retinopathy. <http://atlas.iapb.org/vision-trends/diabetic-retinopathy/> [last accessed March 2018];
5. Flaxman SR, et al. Lancet Glob Health. 2017;5:e1221–34

Motivation



- Detect the changes between the healthy and unhealthy retina
- Overcome the need of a trained opthamologist
- Overcome the complications of DR



- **Complications of DR**
- Vitreous hemorrhage
- Retinal detachment
- Glaucoma
- Blindness

**Build a Deep learning
model for DR Image
Classification**

**Facilitate the DR
Screening Process**

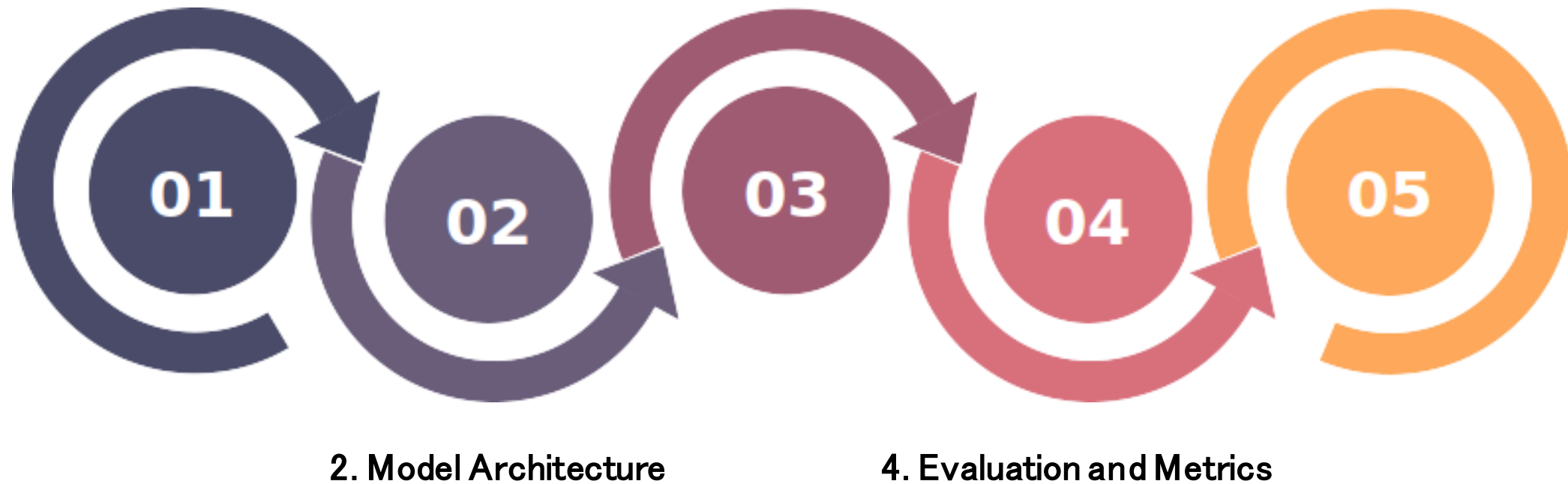
**Creating an aid for
medical diagnostics
in the field of DR**



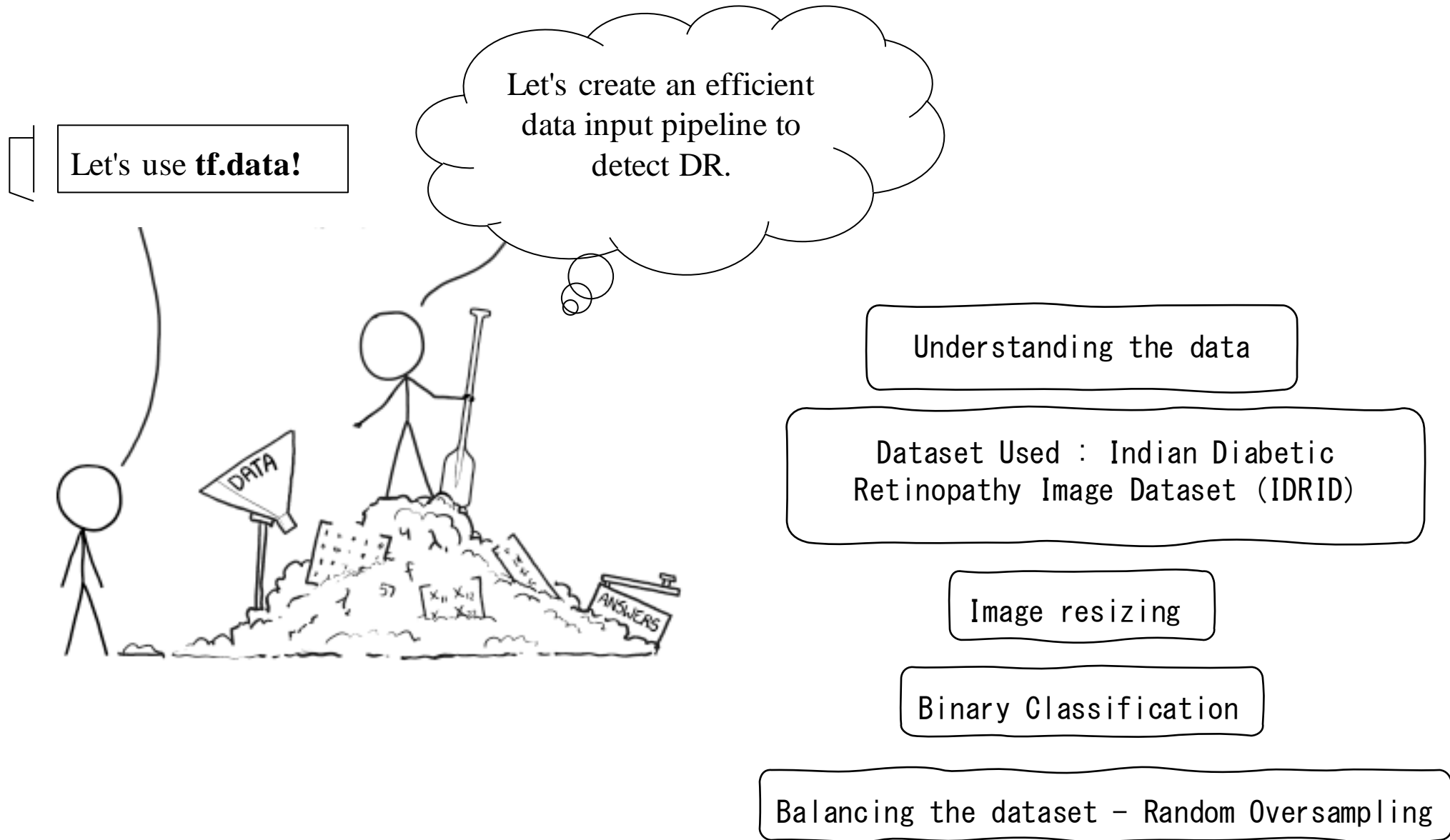
1. Data Preprocessing –
Input Pipeline

3. Model Training

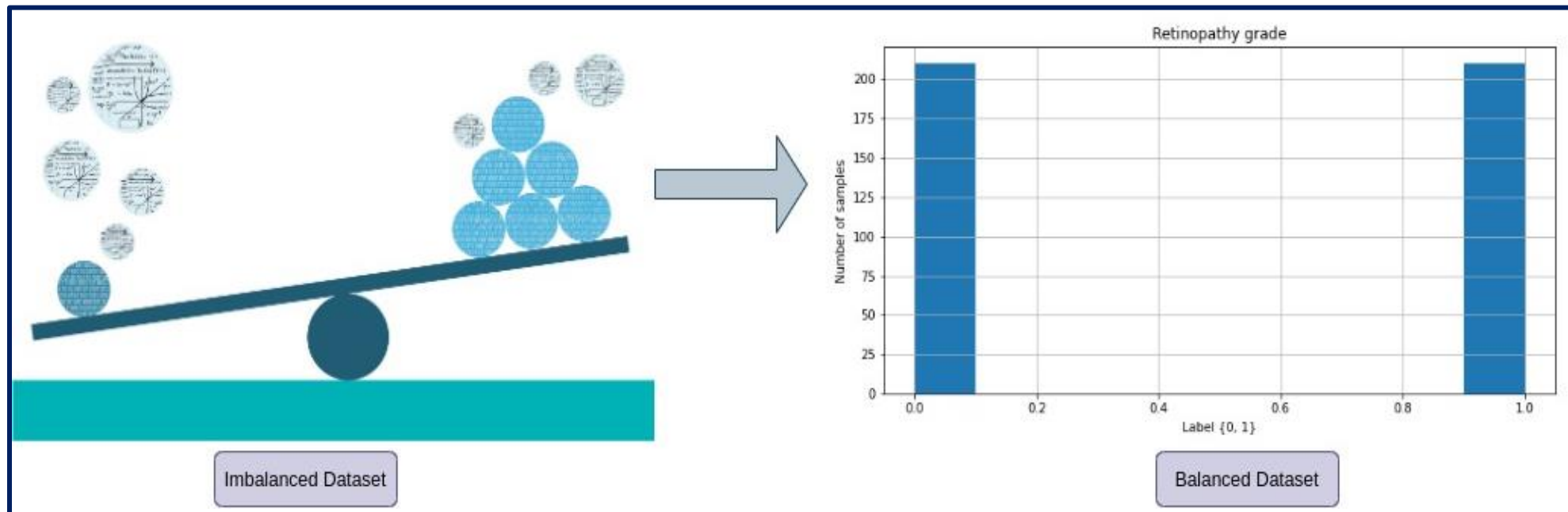
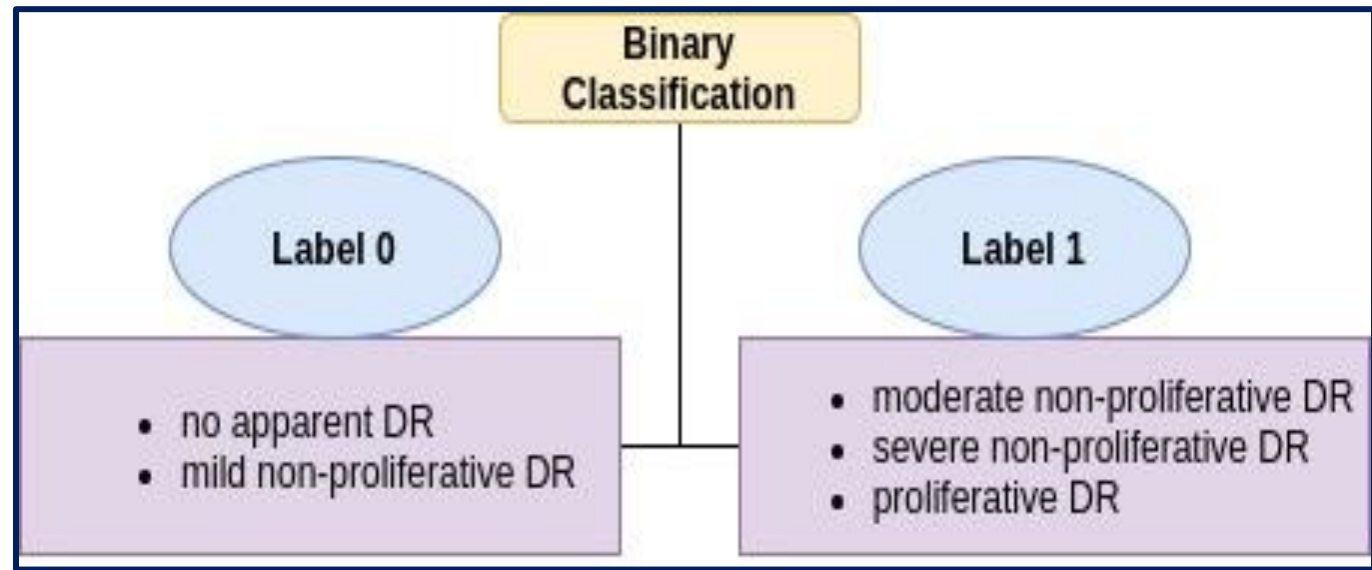
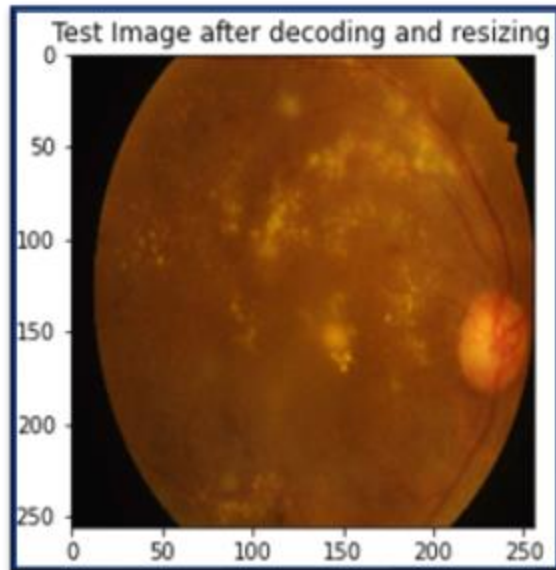
5. Deep Visualization



Data Preprocessing - Input Pipeline



Data Preprocessing - Input Pipeline



Input Pipeline Creation



516 - Fundus Images

413 – Train Images

103 – Test Images

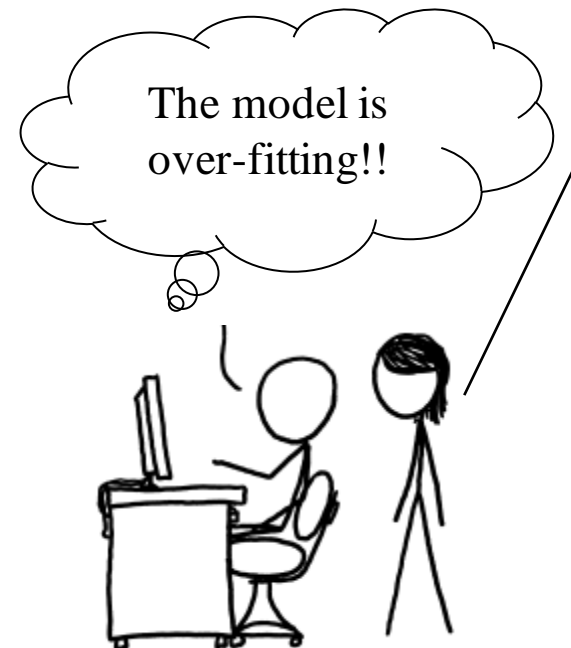
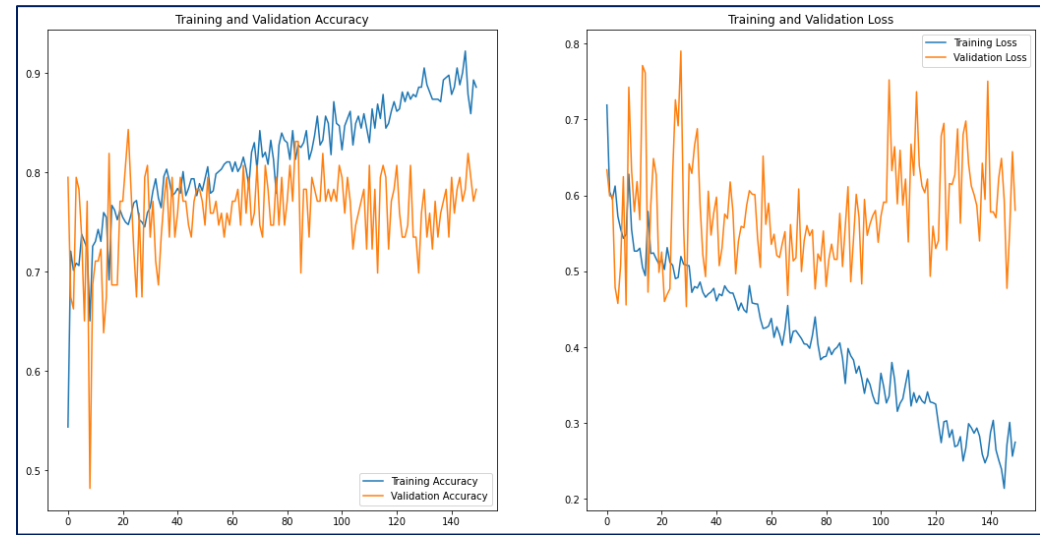
Input image size – 256 x 256

Configuring dataset for performance

To train a model precisely, the following are to be taken care of,

- Shuffling
- Batching

Model with overfitting

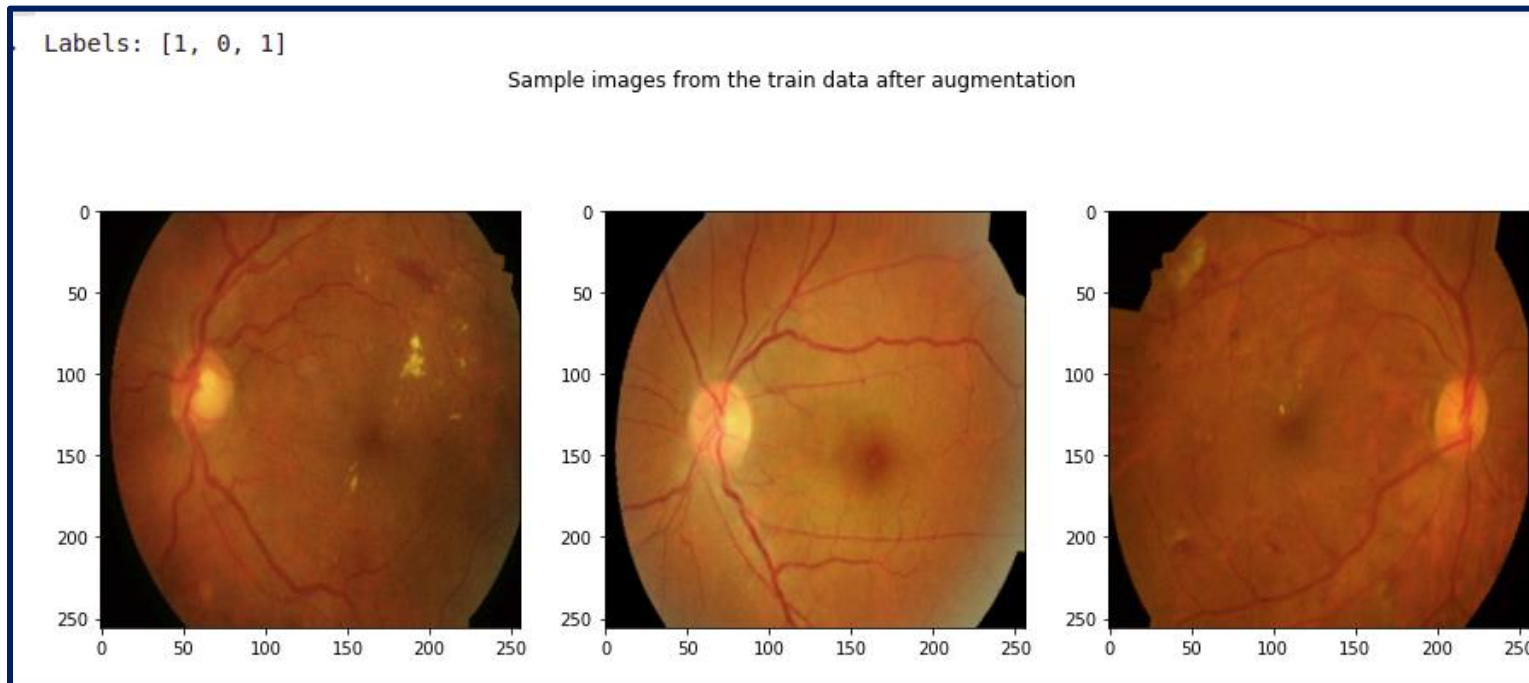


Well! We can try augmenting the images. Also take a note of the dropout layer too!

1. Data Augmentation

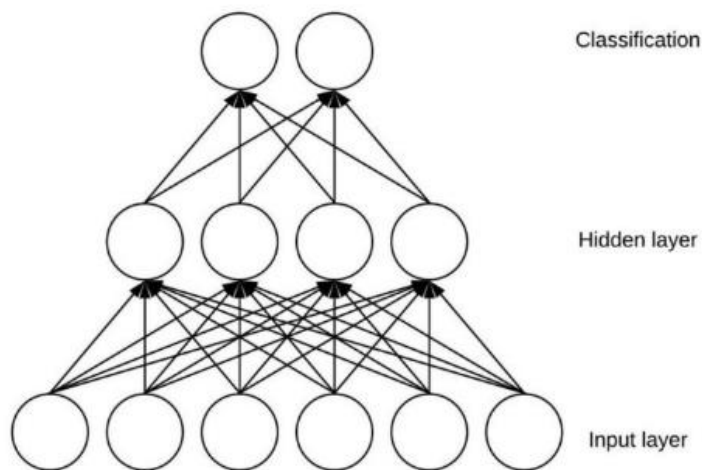
Operations Performed

- Rotation
- Zoom
- Shift
- Horizontal and Vertical Flipping

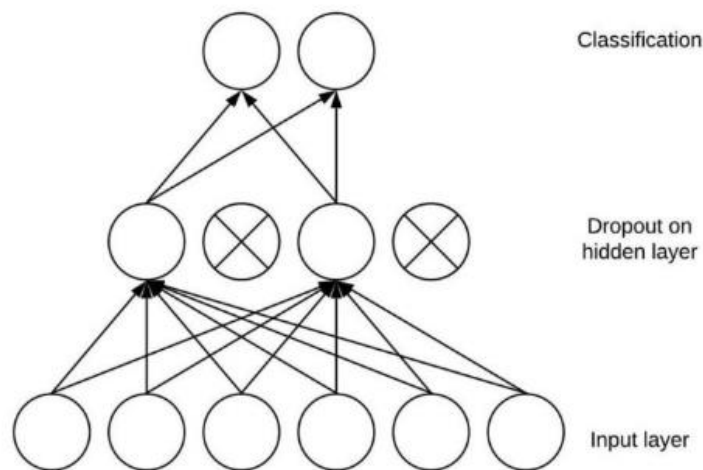


2. Adding Dropout Layers

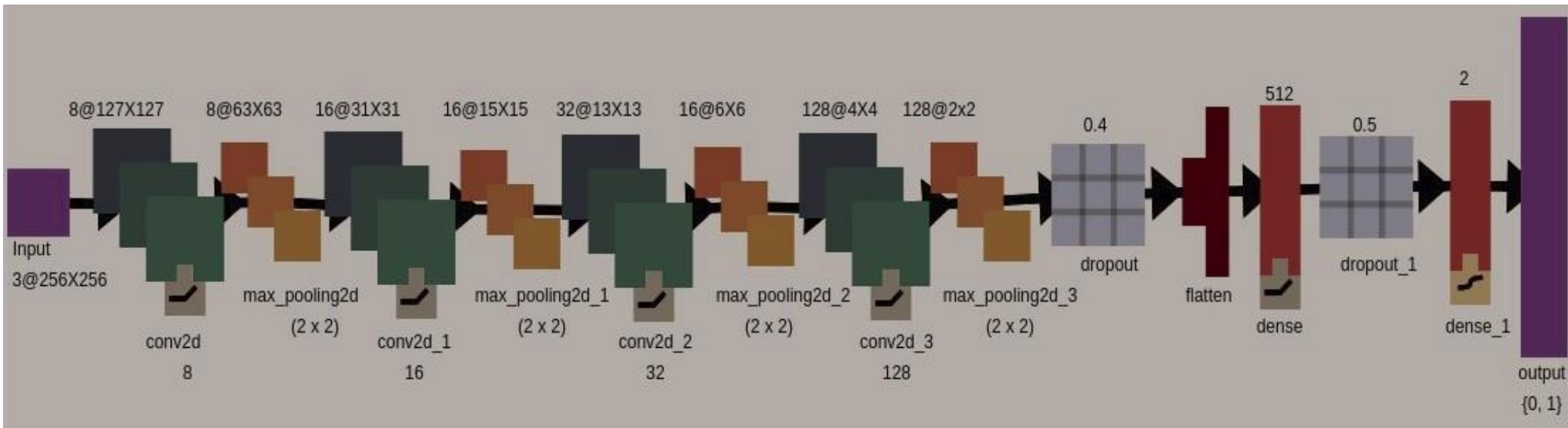
Input of Dropout Layer	Fractional Number (0.1, 0.2, 0.3,.....)
Action Performed	Drops 10%, 20%, 30%, of the output units randomly from the applied layer.



Without Dropout

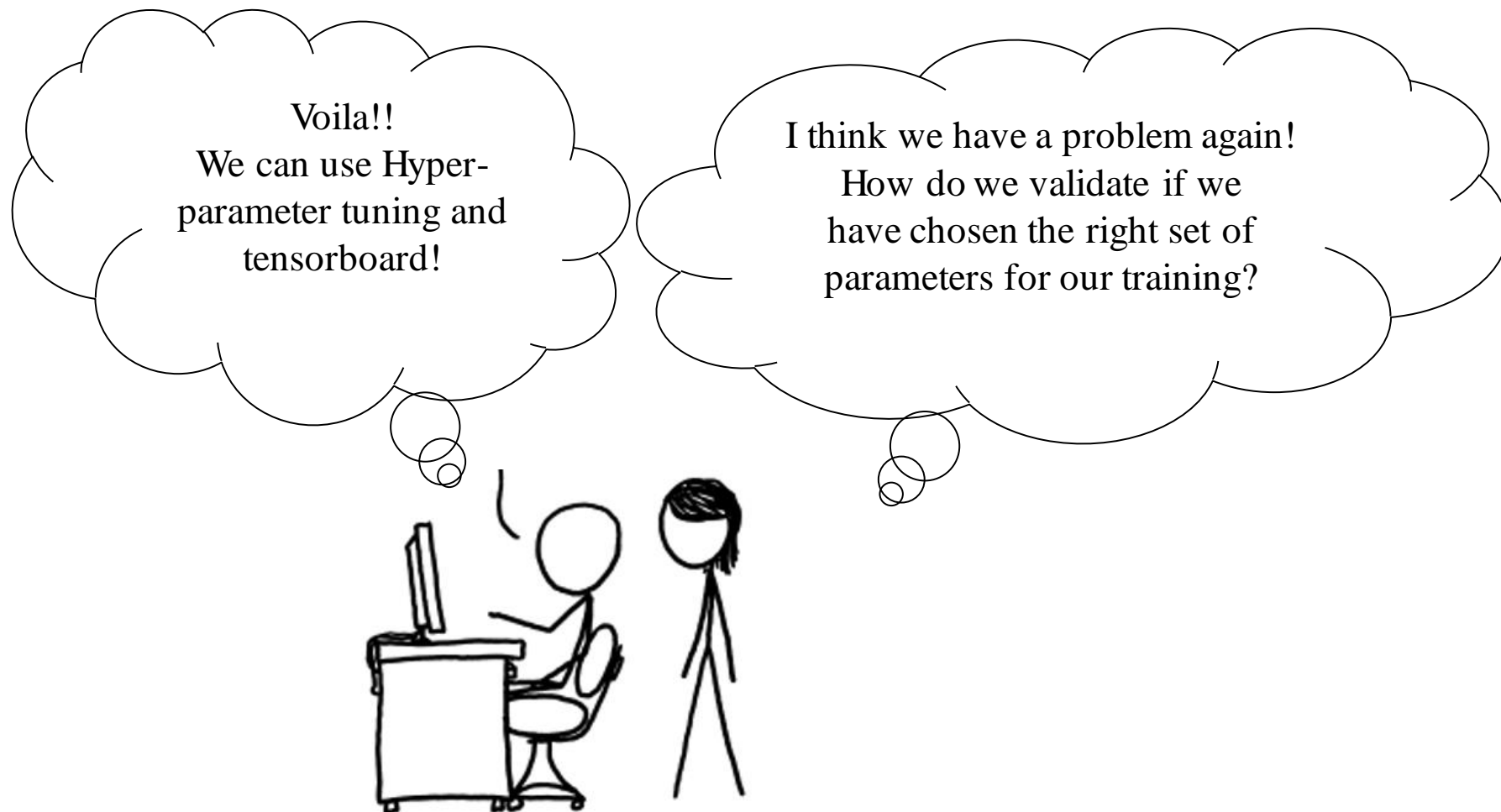


With Dropout



The model consists of,

- Convolutional Neural Network(CNN)
- Max-Pooling Layer
- Dropout



Hyperparameter Tuning (HPT)

- HP_OPTIMIZER
- HP_EPOCHS
- HP_DENSE_LAYER
- HP_DROPOUT

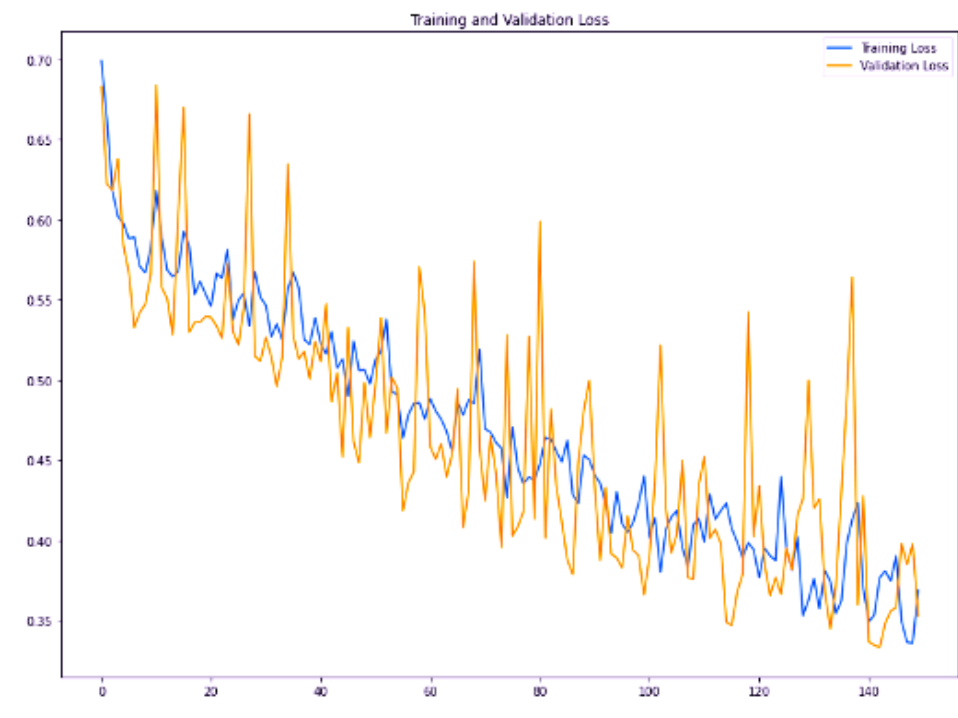
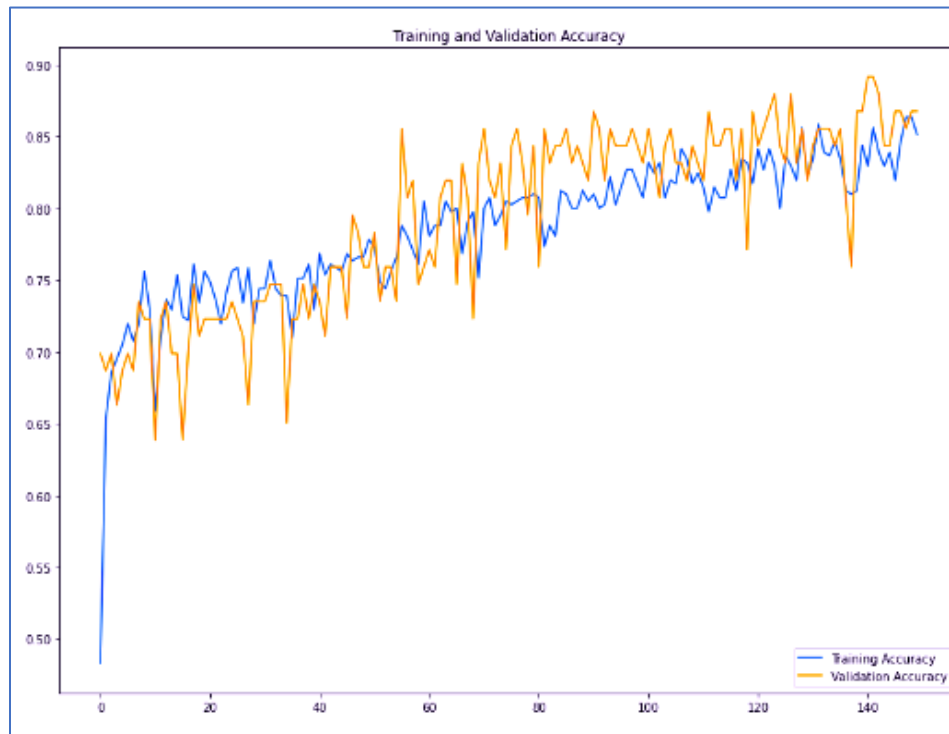
Selection based on HPT

Optimizer - Adam

Number of Epochs - 150

Dense layer - 512

Dropout - 0.4, 0.5



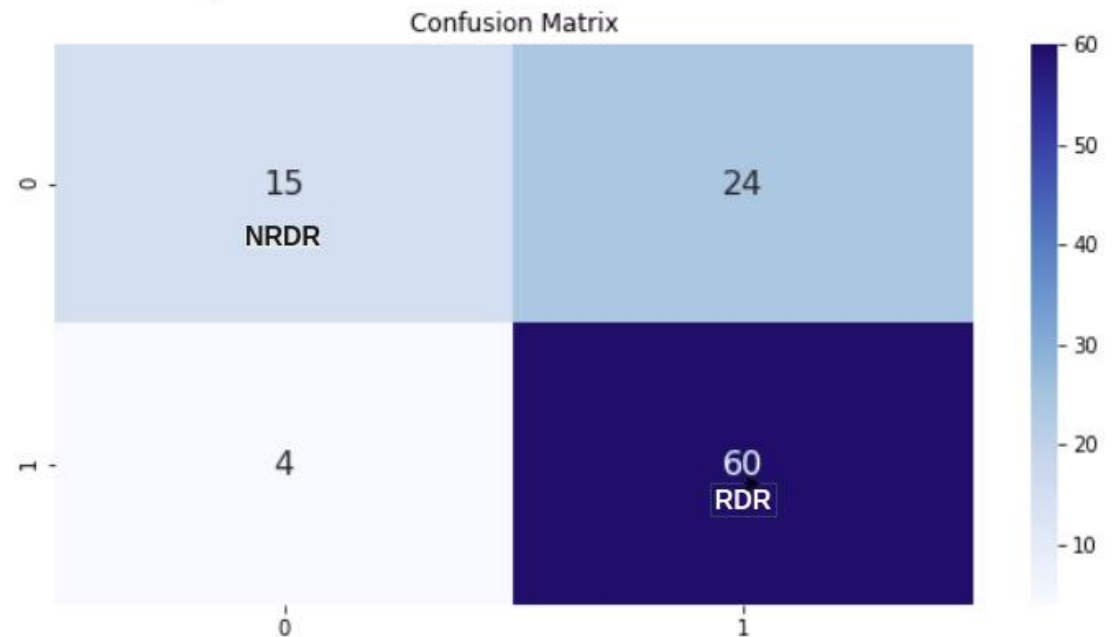
Train and Validation Accuracy and Loss

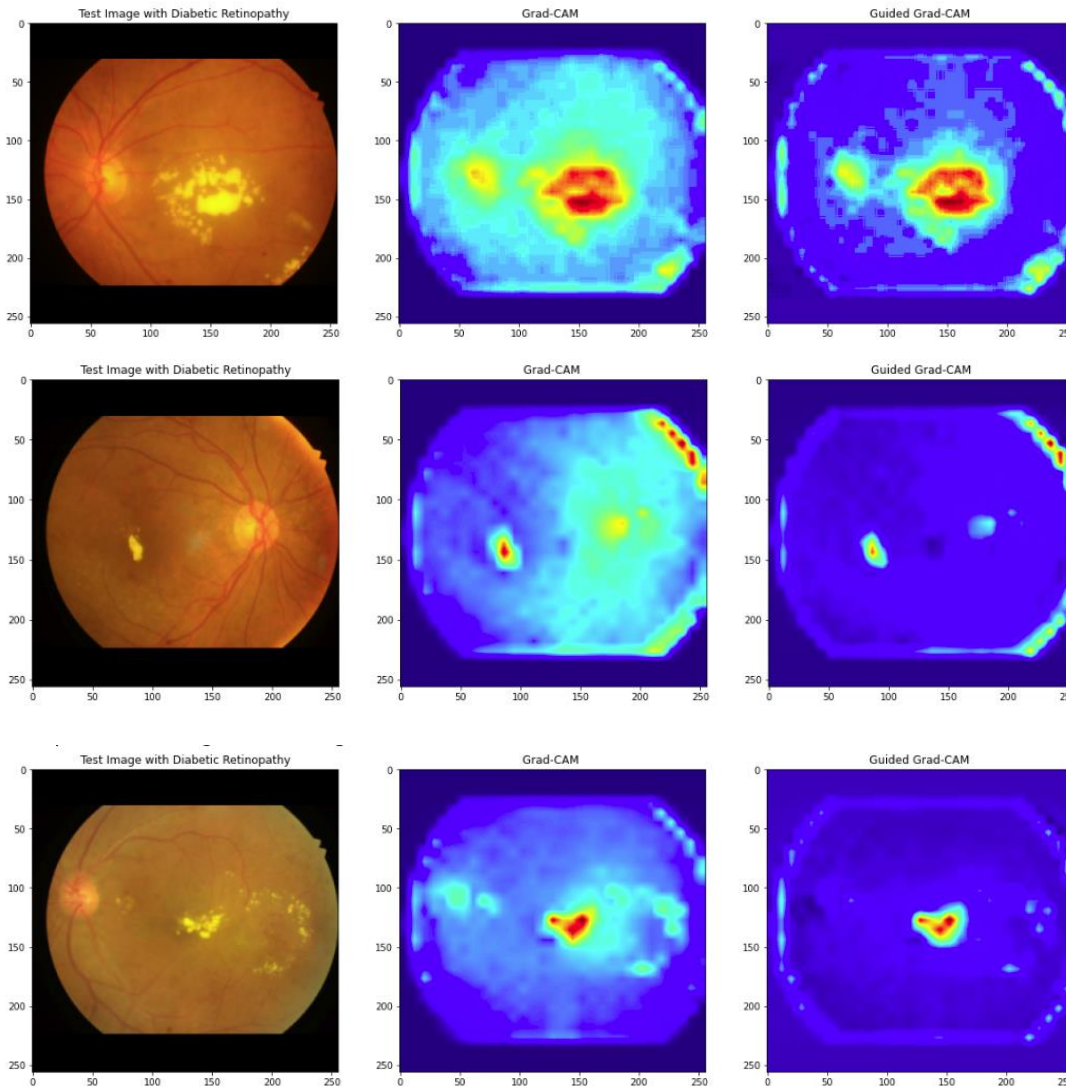
Metrics

Confusion Matrix : Describes the performance of the model
The final achieved test accuracy is about **72.81%**.

		Predicted class	
		<i>P</i>	<i>N</i>
Actual Class	<i>P</i>	True Positives (TP)	False Negatives (FN)
	<i>N</i>	False Positives (FP)	True Negatives (TN)

Test Accuracy: 0.7281553398058253





Grad-Cam

- Localizes image region with relevance
- Visualizes gradients of the final conv2D layer

Guided Backpropagation

- Nullifies the gradients associated with negative value

Conclusion and Future Scope

- Model for early detection of patients with the highest risk of vision loss.
- Timely referral to a retinal specialist
- Aims at preventing irreversible vision impairments.
- Feasible deployment across different eye-care centers

Future Scope:

- Diabetics can be classified as , Type1 – Juvenile and Type 2 – Elderly Individuals
- Dataset can be collected and analysed based on the,
- Age | Type of Diabetics
- Drug used – Insulin and Oral

References

- [1] https://www.tensorflow.org/tutorials/load_data/images
- [2] Selvaraju, Ramprasaath R. ; Cogswell, Michael ; Das, Abhishek ; Vedantam, Ramakrishna ; Parikh, Devi ; Batra, Dhruv: Grad-cam: Visual explanations from deep networks via gradient-based localization. In: Proceeding Conference on Computer Vision, 2017, S. 618–626
- [3] <https://math.mit.edu/ennui/>
- [4] Deep Learning Course and Lab lectures and slides

Thank you