

# Diagnosing diabetic retinopathy from images of the eye fundus

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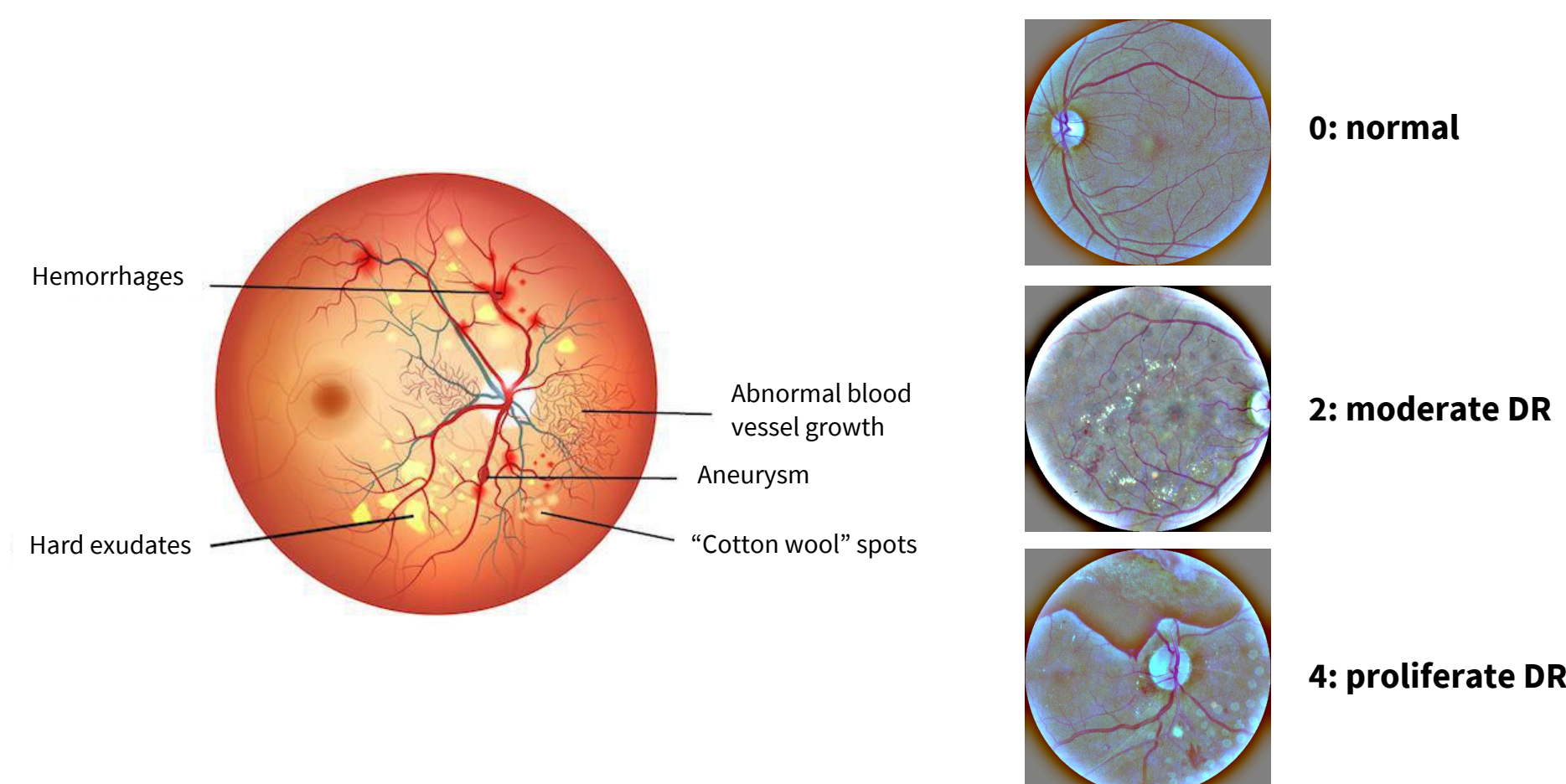
<https://youtu.be/PzuAppZm9R8>

## Introduction

- Diabetic retinopathy (DR) is leading cause of blindness in working aged adults
- Need for regular screening without specialist in rural areas
- Use convolutional neural network to classify eye fundus images into 5 disease stages

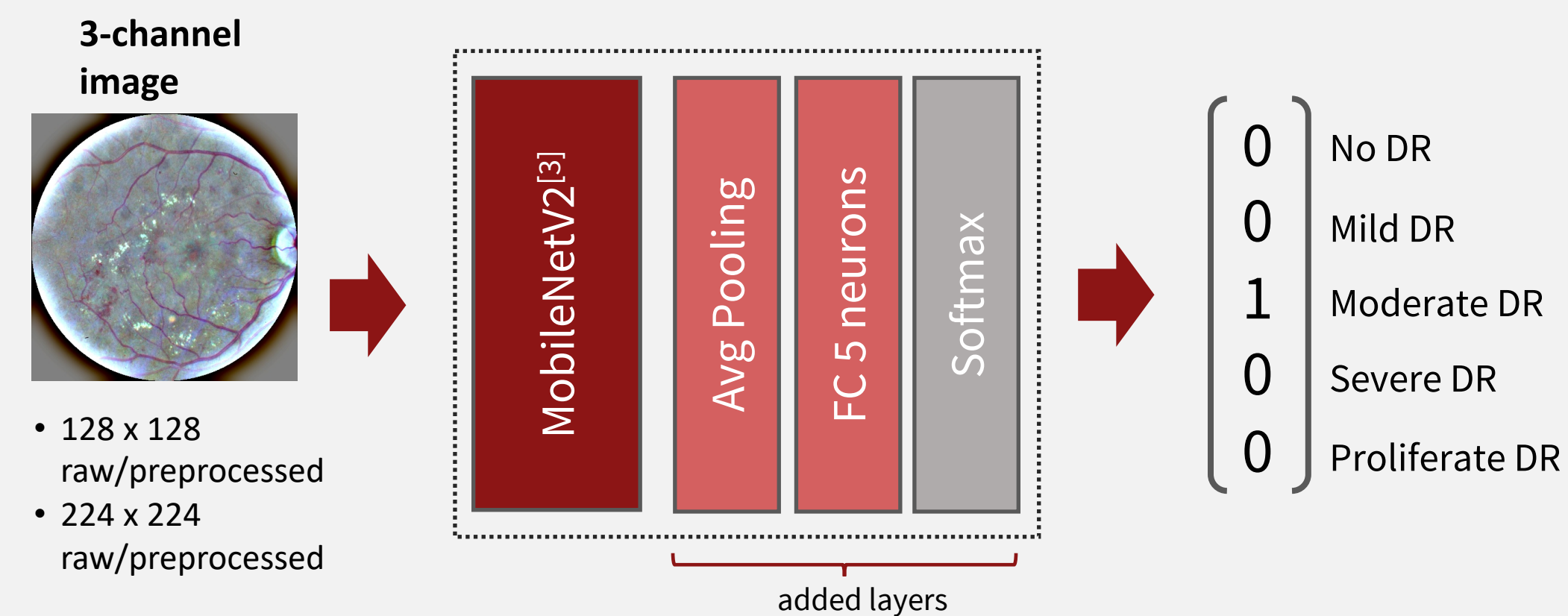
## Dataset

- 3662 labeled 3-channel images from APTOS 2019 Blindness detection challenge on Kaggle [1]
- 5 classes for different disease stages
- Imbalanced class prevalence
- Split into train/dev/test: 70/15/15 preserving class distribution



[2]

## Model



## Methods

- Categorical cross entropy loss
- Adam optimizer
- Categorical accuracy
- Quadr. Weighted kappa (QWKP)

## Model optimization

- Learning rate
- image size, image preprocessing [4]
- Unfreeze weights
- Weighted loss
- Data augmentation

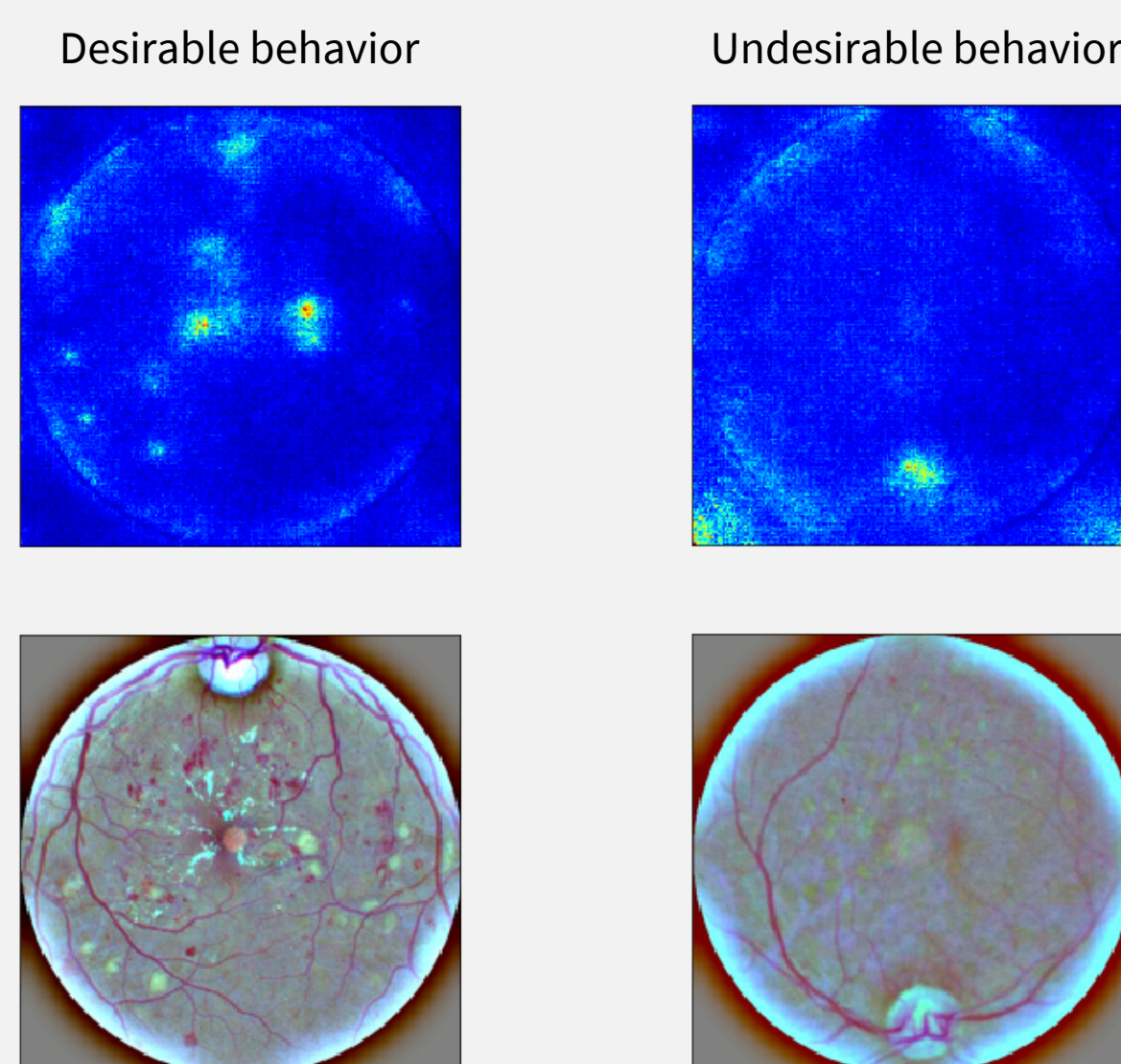
## Results

	Categorical accuracy Training set		Categorical accuracy Development set	
<b>Initial</b>	LR 0.001	0.8246	LR 0.001	0.4926
Raw images 224x224	LR 0.0001	0.7267	LR 0.0001	0.6158
Model weights frozen	LR 0.00001	0.6580	LR 0.00001	0.487
<b>Unfreeze model weights</b>			Accuracy	0.6324
Raw images 224x224		0.9767	QWKP	0.418
LR: 0.0001				
<b>Weighted loss</b>			Accuracy	0.5827
Raw images 224x224		0.9875	QWKP	0.6833
LR: 0.0001				
<b>Data augmentation</b>			Accuracy	<b>0.6930</b>
Rotation, flipping			QWKP	0.7761
Preprocessed 224x224		<b>0.8066</b>	<b>Test set</b>	
LR: 0.0001			Accuracy	<b>0.7847</b>
Weighted loss			QWKP	0.8123

- Initially high bias and variance
- Unfreezing the pretrained model weights reduces bias
- Data augmentation reduces variance
- Variance still noticeable

## Model interpretation

### Saliency maps



## Conclusion

- Challenges: small amount of data, class imbalance
- Strategies: transfer learning, data augmentation, weighted loss
- Performance comparable to lower field of kaggle competitors

## Future work

- Cropping background
- L2 regularization
- More data augmentation
- Deeper analysis of saliency maps and class activation maps
- Collect more data

## References

[1] <https://www.kaggle.com/c/aptos2019-blindness-detection>  
 [2] <https://www.eyepops.com/contents/our-services/eye-diseases/diabetic-retinopathy>

[3] Sandler, M., Howard, A., Zhu, M., Zhmoginov, A. and Chen, L.C., 2018. Mobilenetv2: Inverted residuals and linear bottlenecks. Proceedings of the IEEE conference on computer vision and pattern recognition, pp. 4510-4520.

[4] <https://www.kaggle.com/matheuseduardo/preprocessed-2019-blindness-detection>