

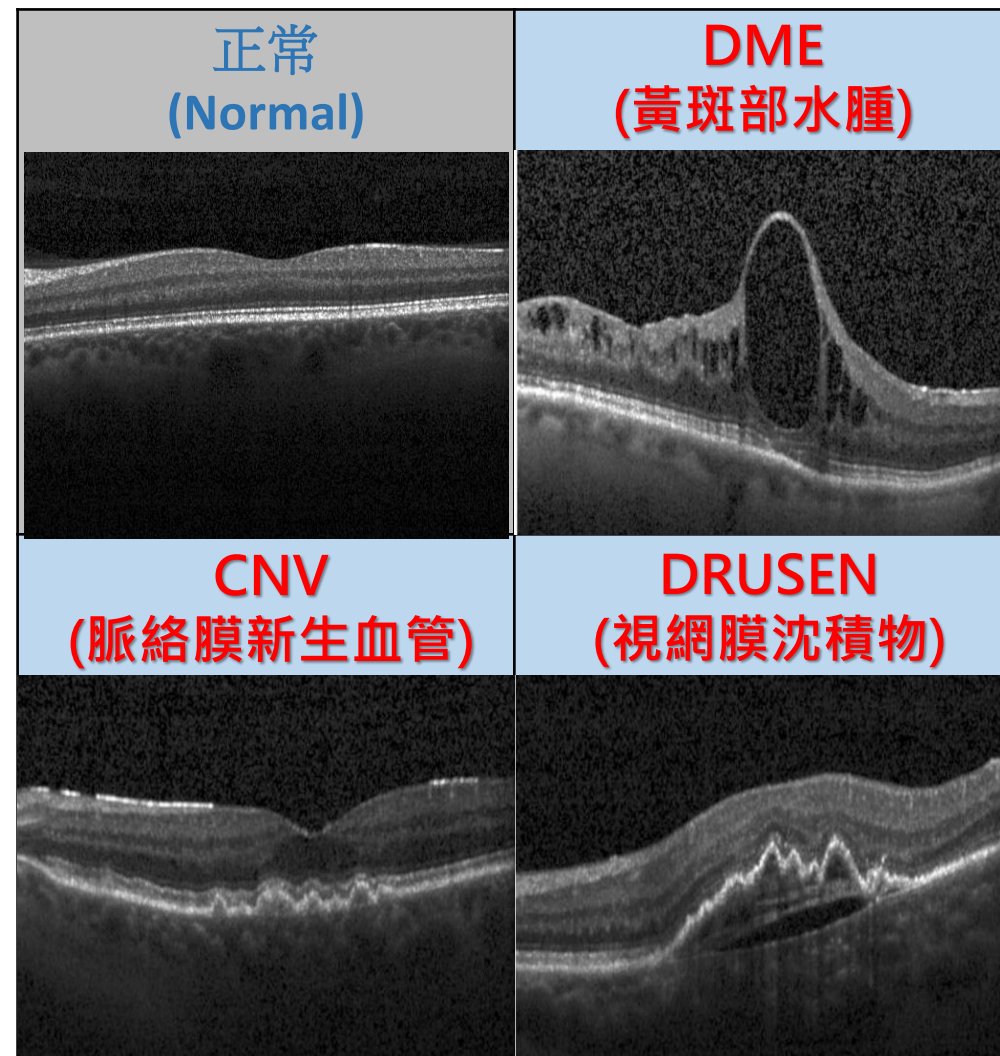
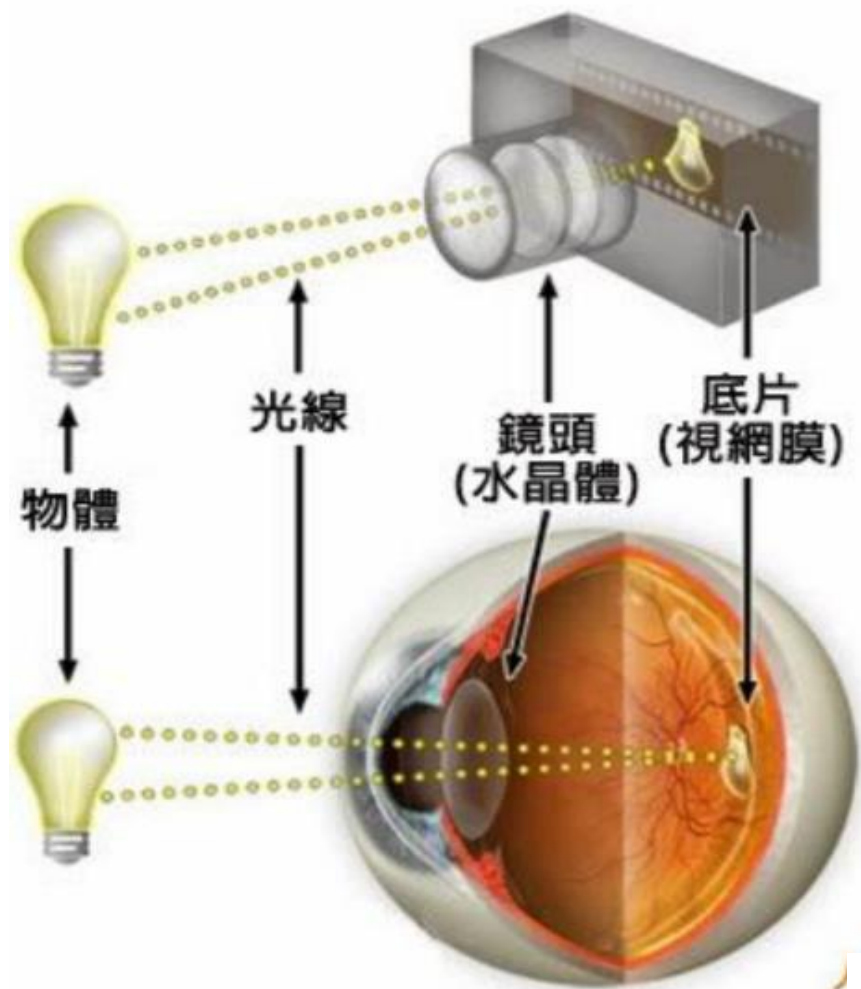


# Detecting Retina Symptom

## 視網膜症狀檢測

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# 視網膜介紹與症狀





# 視網膜檢測時有以下問題：



誤判

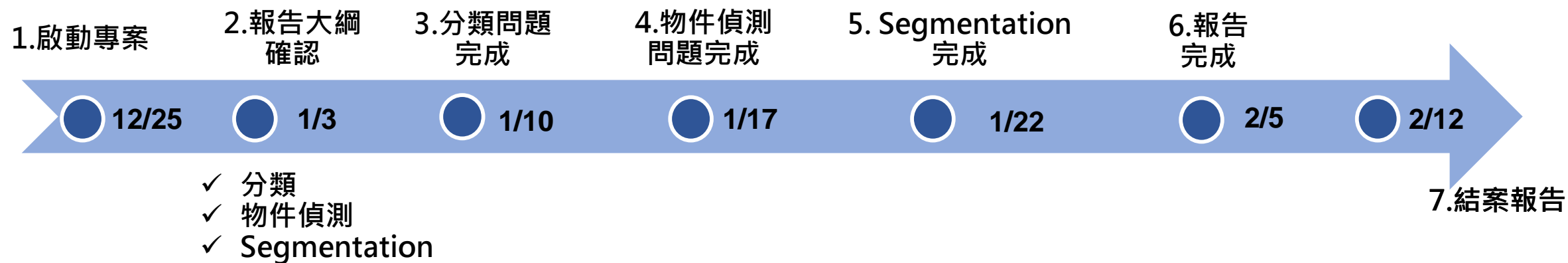


耗時

本專案期望以AI 技術輔助檢測，降低醫生診斷時負擔



# 解題過程



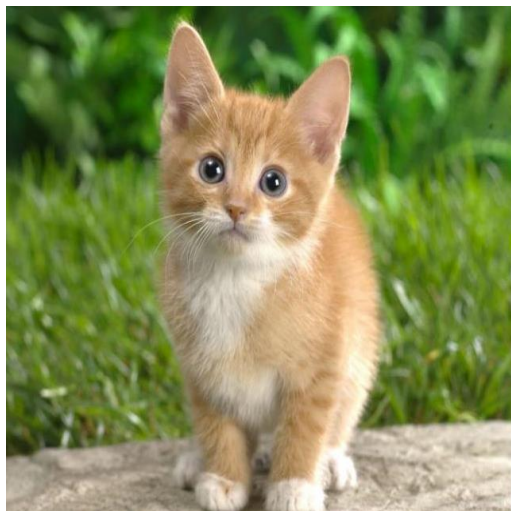
# 視網膜檢測攻堅思維

有沒有症狀？

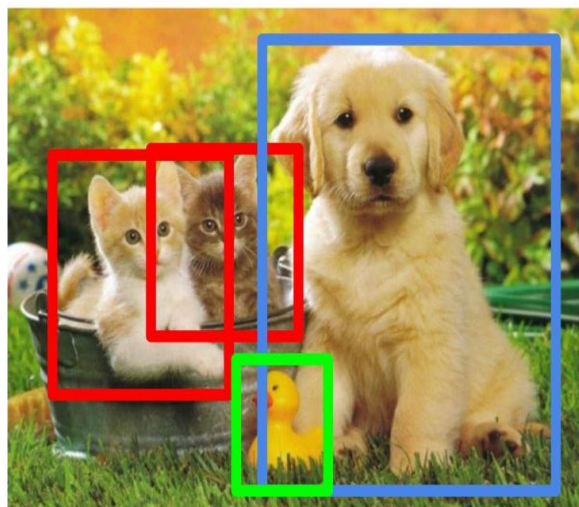
什麼症狀？

發生在那/多大？

物件歸類  
Classification



物件偵測  
Object Detection



語意分析  
Instance Segmentation



本專案以電腦視覺三大方法做為主要攻堅計劃

# 執行 AI 專案注意事項

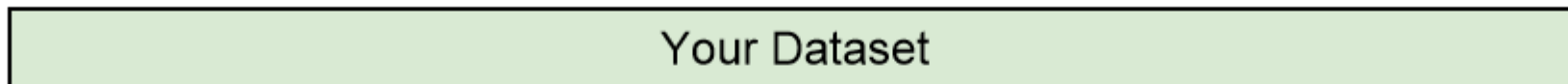
## Tips:

- ✓ 資料拆成 Training / Dev / Test
- ✓ 注意 High Bias and High Variance
- ✓ 注意 Dev set and Test Set same distribution
- ✓ Error Analysis

# ✓ 資料拆成 TRAINING / DEV / TEST

**Idea #1:** Choose hyperparameters that work best on the data

**BAD:** Easy works perfectly on training data



**Idea #2:** Split data into **train** and **test**, choose hyperparameters that work best on test data

**BAD:** No idea how algorithm will perform on new data



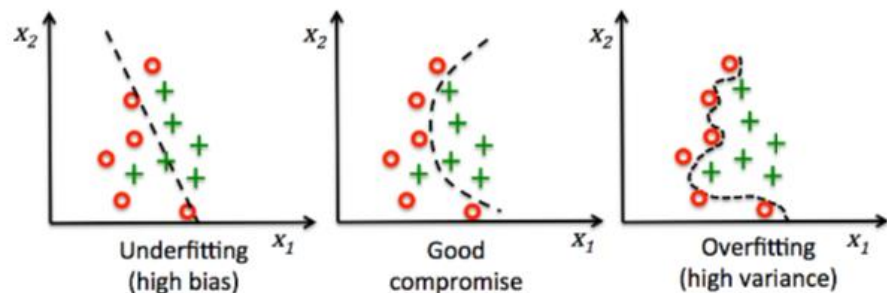
**Idea #3:** Split data into **train**, **val**, and **test**; choose hyperparameters on val and evaluate on test

**Better!**



# ✓ 注意 HIGH BIAS AND HIGH VARIANCE

➤ 何謂 High Bias , High Variance



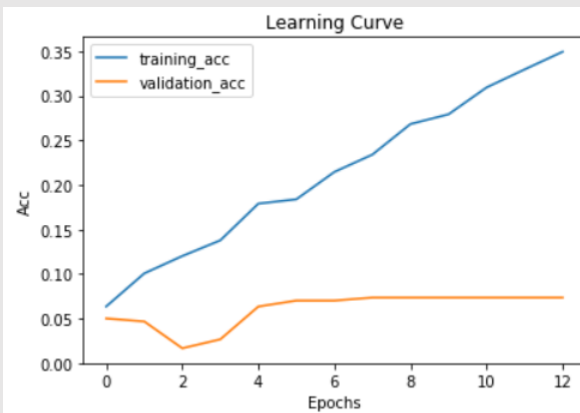
➤ Action of High Bias , High Variance Problems

High Bias	High Variance
1. Train bigger model	1. More data
2. Train longer/better optimization algorithms	2. Regularization
3. NN architecture / hyperparameters search	3. NN architecture / hyperparameters search

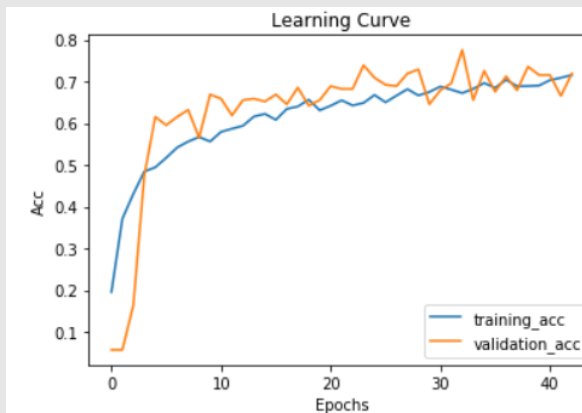
參考資料:  
Coursera , Deep Learning, Structuring Machine Learning Projects, Andrew Ng

本例 human-level performance  $\approx 1$

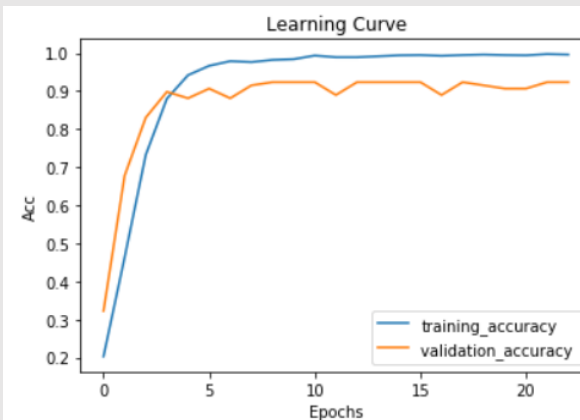
High Bias & High Variance



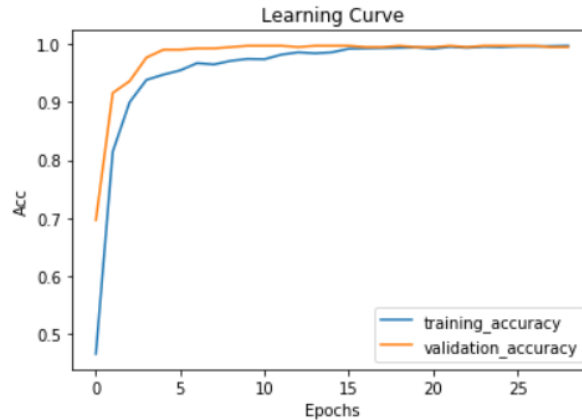
High Bias



High Variance



Perfect





# ✓ 注意 DEV SET AND TEST SET SAME DISTRIBUTION

Regions:

- US
- UK
- Other Europe
- South America
- India
- China
- Other Asia
- Australia

Dev

Test

Randomly shuffle into dev/test



Guideline:

Choose a dev set and test set to reflect data you expect to get in the future and consider important to do well on.

Same Distribution

# ✓ ERROR ANALYSIS

- 製作表格，針對分類錯誤的 Image 下 Comments

Image	Dog	Great Cat	Blurry	Incorrectly labeled	Comments
1				✓	Labeler missed cat in background
2		✓			
3				✓	Drawing of a cat; Not a real cat.
...					



✓ 有了 AI Project Knowledge 後，開始實作

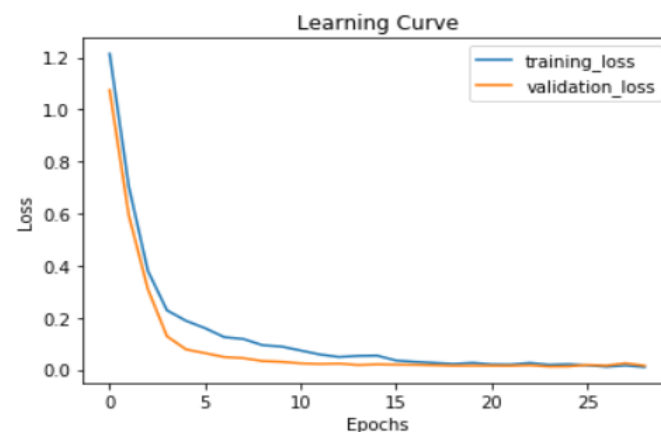
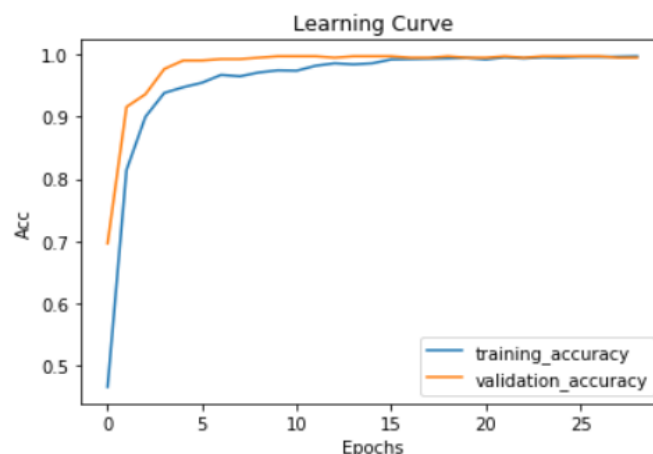
# 攻堅計劃一：影像分類判定症狀

## ● Basic Information

Method		Transfer Learning
Model		Xception
Hyperparameter	optimizer	Adam(lr=10e-6)
	epochs	100
ImageDataGenerator		rotation_range=10, width_shift_range=0.1, height_shift_range=0.1, shear_range=0.1, zoom_range=0.1, horizontal_flip=True, fill_mode='nearest')
Earlystop		val_loss 連續 5 epoch 無下降
Data		CNV 605 DME 501 DRUSEN 580 Normal 517
Train_Test_Split		0.8/0.2

## ● Training/Validation Accuracy/Loss

- Model 無 bias · variance 問題



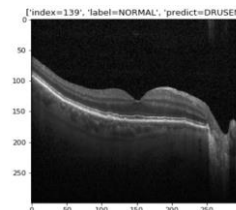
- 錯誤分析: 模型預測 Validation Set 結果  
正常照片 recall 0.99/異常照片 recall 1  
=> Model 無漏篩問題

predict	CNV	DME	DRUSEN	NORMAL
label				
CNV	117	0	0	0
DME	0	87	0	0
DRUSEN	0	0	112	0
NORMAL	0	0	1	124

Validation loss: 0.011931998532713135

Validation accuracy: 0.9977324

	precision	recall	f1-score	support
CNV	1.00	1.00	1.00	
DME	1.00	1.00	1.00	
DRUSEN	0.99	1.00	1.00	
NORMAL	1.00	0.99	1.00	
accuracy			1.00	441
macro avg	1.00	1.00	1.00	441
weighted avg	1.00	1.00	1.00	441





# 攻堅計劃一：影像分類成效驗證

- 新資料預測 (每類各 100 張)

正常照片 recall 為 1 (100/100)

異常照片 recall 為 0.987 (4/300)

=> Model 無漏篩問題

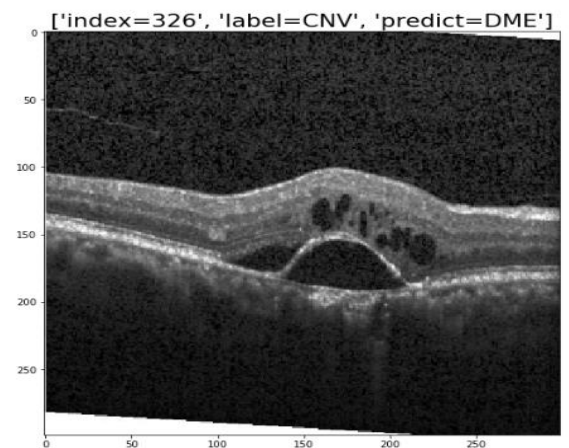
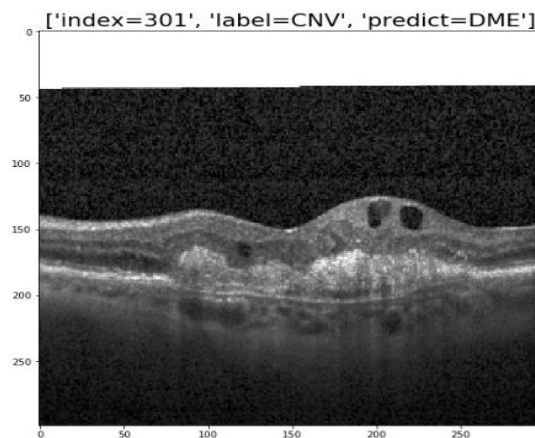
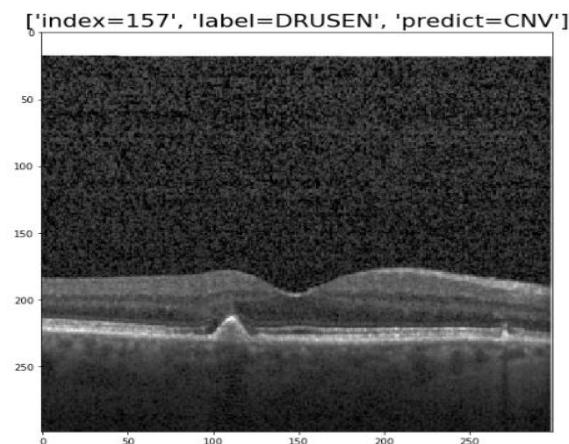
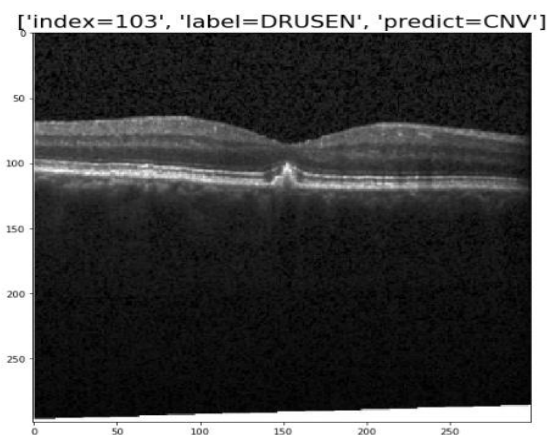
predict \ label	CNV	DME	DRUSEN	NORMAL
CNV	98	2	0	0
DME	0	100	0	0
DRUSEN	2	0	98	0
NORMAL	0	0	0	100

label \ predict	CNV	DME	DRUSEN	NORMAL
CNV	98	2	0	0
DME	0	100	0	0
DRUSEN	2	0	98	0
NORMAL	0	0	0	100

	precision	recall	f1-score	support
CNV	0.98	0.98	0.98	100
DME	0.98	1.00	0.99	100
DRUSEN	1.00	0.98	0.99	100
NORMAL	1.00	1.00	1.00	100
accuracy			0.99	400
macro avg	0.99	0.99	0.99	400
weighted avg	0.99	0.99	0.99	400

- 失效影像驗證:

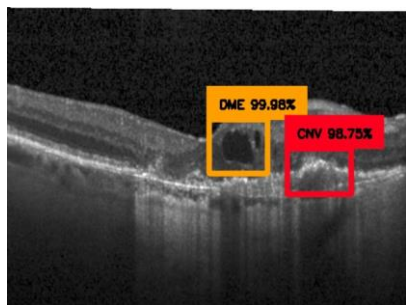
複合型異常，導致分類錯誤。解法: Object Detection



# 攻堅計劃二：物件偵測解決複合型異常

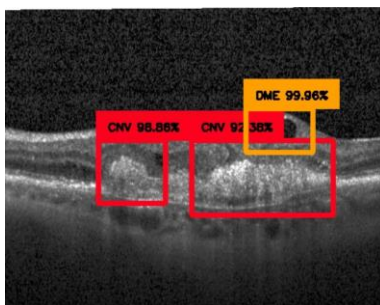
## ● Basic Information

Method	YOLOv3	
Hyperparameter	epochs	30
Number of label	CNV 252 DME 146 DRUSEN 183	



### 準確度

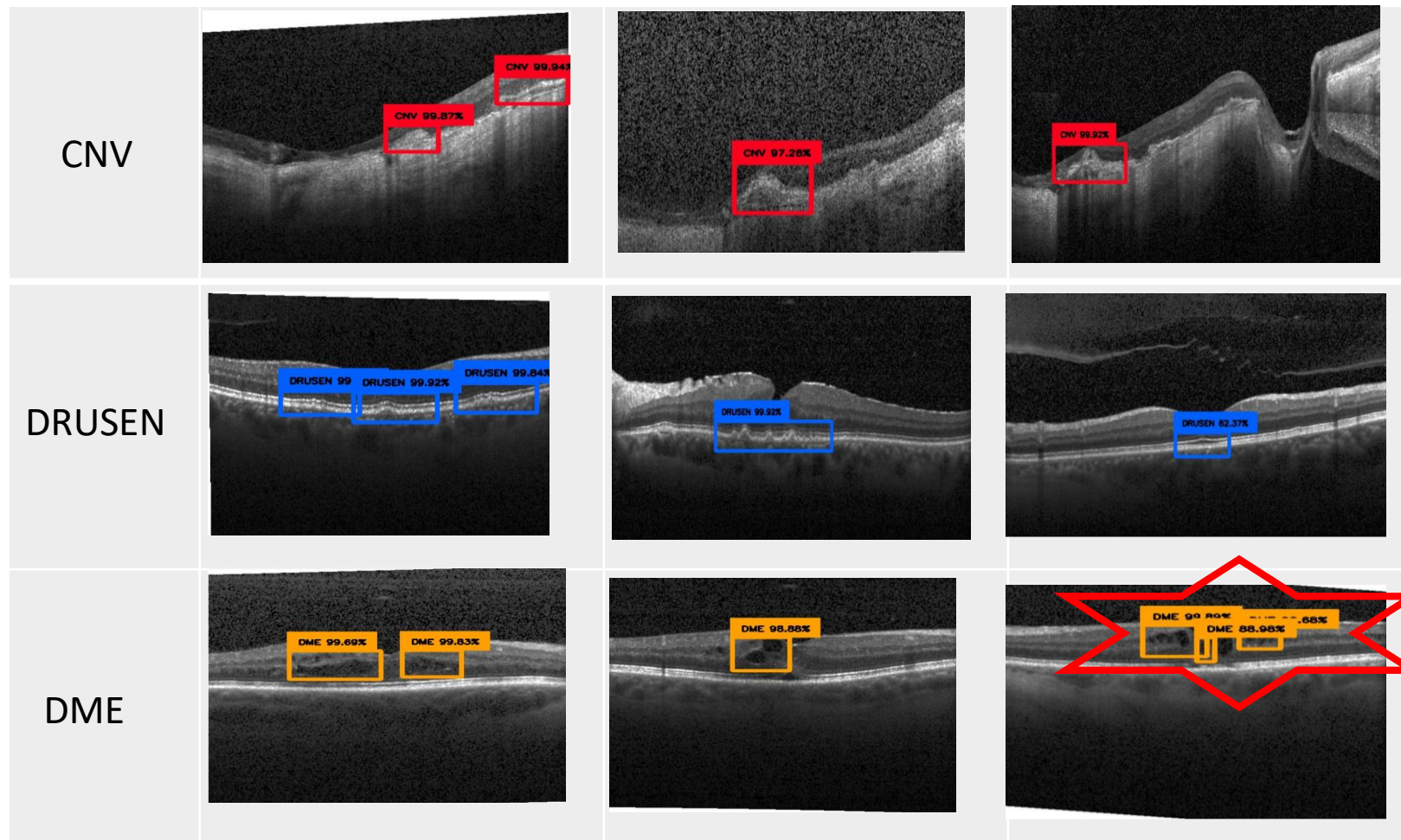
CNV: 0.9020  
DME: 0.9540  
DRUSEN: 0.9279  
mAP: 0.9280



### 平均準確度

結果: mAP: 0.928

- 成效驗證：平均準確度mAP 達92.8%
- 後續問題：偵測的 Box 集中不美觀。解法: Segmentation

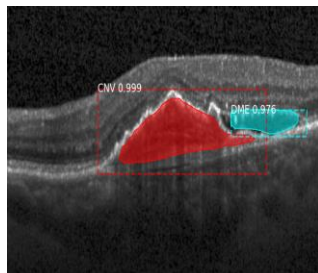
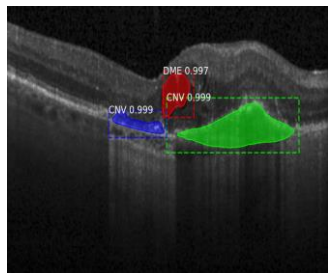




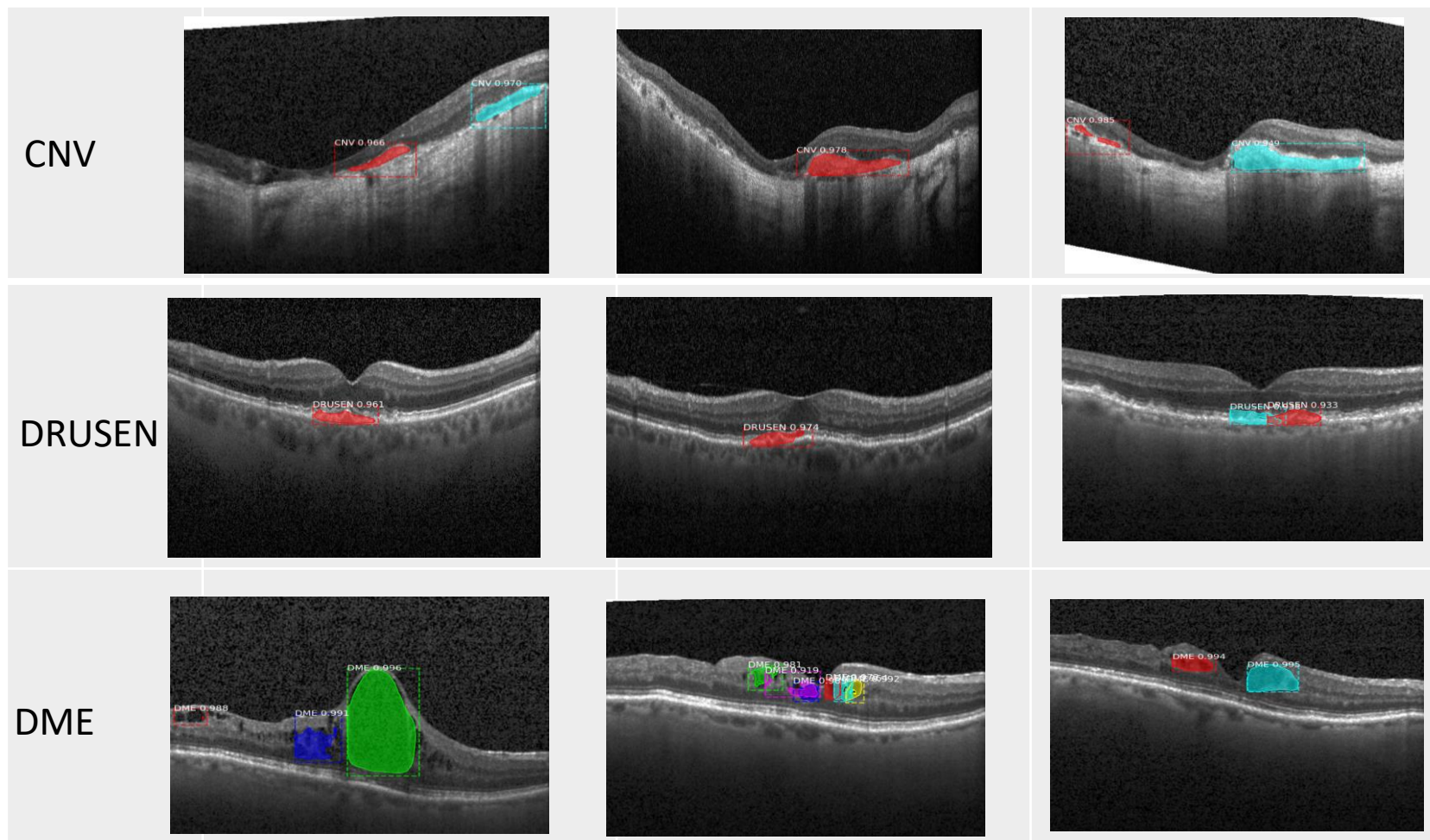
# 攻堅計劃三：語意分析解決偵測BOC 集中問題

## ● Basic Information

Method	Mask R-CNN	
Hyperparameter	epochs	10
STEPS_PER_EPOCH		100
Detection min confidence		0.9
Number of label	CNV	30
	DME	30
	DRUSEN	30

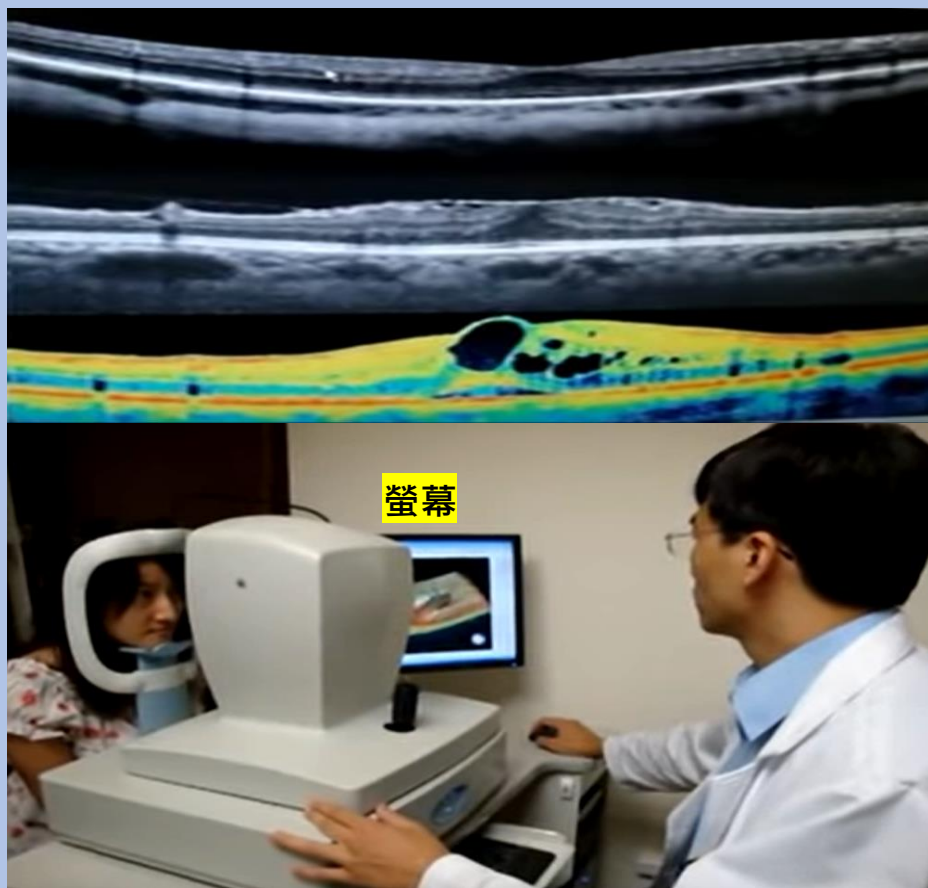


## ● 成效驗證：有效解決物件偵測BOS 集中，不美觀問題 並與物件偵測手法比較



# AI 導入前

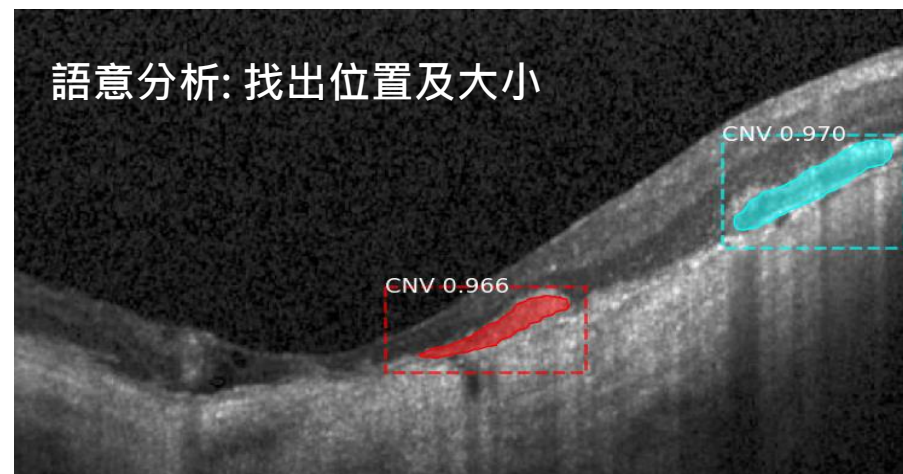
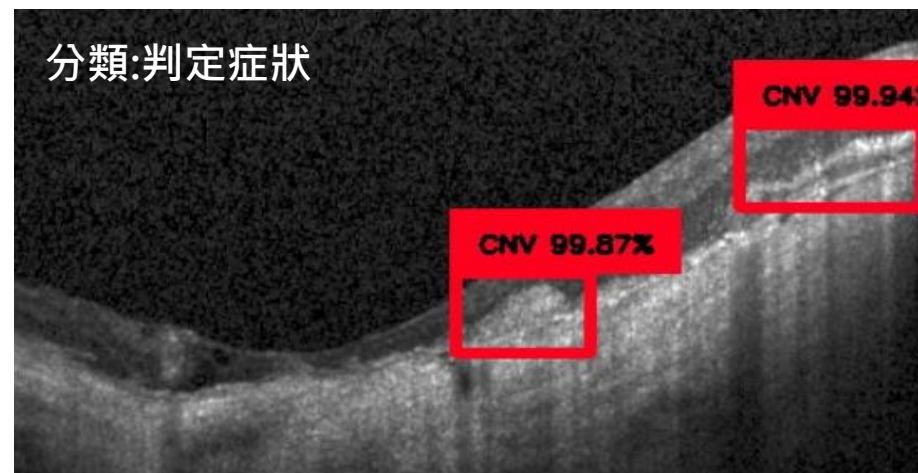
醫生盯著螢幕看症狀



資料來源：  
高雄榮總台南分院眼科李尹暘醫師

# AI 導入後

分類與偵測顯示螢幕輔助判斷





對企業/產業的IMPACT

AI 時代  
一個打十個已  
經不流行了





# 本報告 *Demo*

異常檢測的 AI 專案應有的流程與技術，供各企業參考!!

# 參考資料

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- Coursera , Deep Learning, Structuring Machine Learning Projects, Andrew Ng
- Stanford University CS231n
- AI 學校技術班 YOLOv3 教材
- ORAI Mask R-CNN
- 高雄榮總台南分院眼科李尹暘醫師:視網膜眼斷層掃描 Retina OCT  
<https://www.youtube.com/watch?v=T2kuA5ZfKL4>