

- 分別 LAB 3차 정기 미팅 -
인공지능 기법을 활용한 알츠하이머
Amyloid-PET 분류 정확도 개선

한국인공지능연구소 분별 LAB
김웅곤 연구원

Who am I?



現) 한국인공지능연구소 분별LAB 연구원(2018.6~)

現) 통계청 직원(2016.7~)

서강대학교 수학, 경영학 학사(2012~2017)

고교 수학강사, 대학생 전공수학, 통계 강의 경험 多

API기반 크롤러 개발 배포 & RL 프로젝트 수행 경험 多

목차

1. 중간 발표 결과

2. 개선 Idea

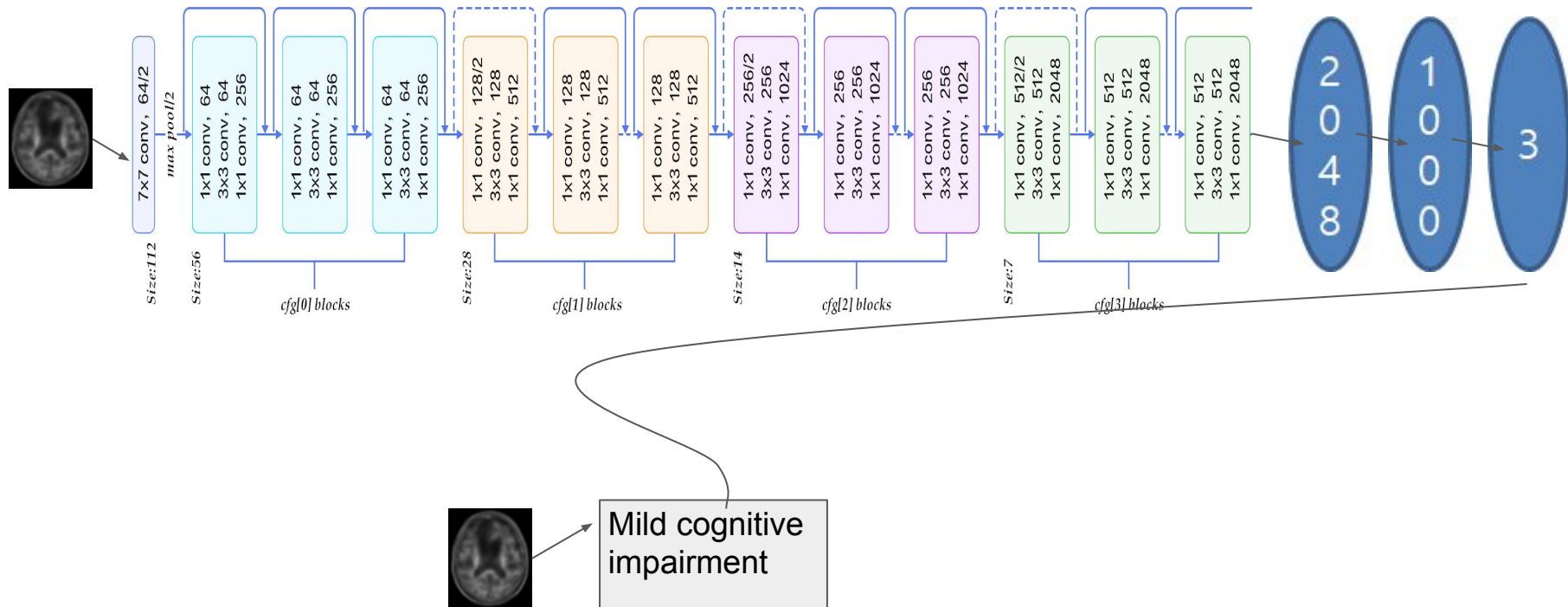
3. 개선 결과

분별 LAB 중간 발표

링크 :

https://docs.google.com/presentation/d/1iUJsk2KJpZ5yU6l1wnP-LfflTkVJ6kHtnkRvo32xDY/edit#slide=id.g3d222f8f52_1_14

LAB 중간발표 - ResNet 50 Model



분별 LAB 중간발표 - The classification report



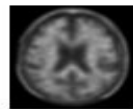
True: 1gr
Pred: 3gr



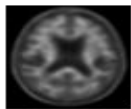
True: 1gr
Pred: 3gr



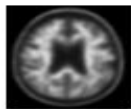
True: 1gr
Pred: 3gr



True: 1gr
Pred: 3gr



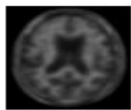
True: 1gr
Pred: 3gr



True: 1gr
Pred: 3gr



True: 1gr
Pred: 3gr



True: 2gr
Pred: 3gr



True: 2gr
Pred: 3gr

Confusion matrix:

```
[[66  0  7]
 [ 8  0 12]
 [ 1  2 56]]
```

| | precision | recall | f1-score | support |
|-------------|-----------|--------|----------|---------|
| 0 | 0.88 | 0.90 | 0.89 | 73 |
| 1 | 0.00 | 0.00 | 0.00 | 20 |
| 2 | 0.75 | 0.95 | 0.84 | 59 |
| avg / total | 0.71 | 0.80 | 0.75 | 152 |
| (0) 1gr | | | | |
| (1) 2gr | | | | |
| (2) 3gr | | | | |

F1 score 75 %, Recall 80 %

If Possible, We need an improvement!

개선 IDEA

복잡한 모델인 ResNet50 대신에

VGG16같은 쉬운 모델로 실험해 보자

무작정 돌려보니 잘 되네????

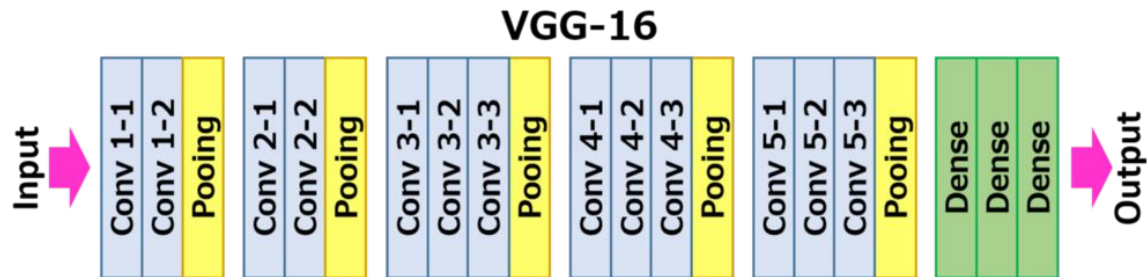
시간 절약 + 남는 시간에 다른 곳에 투자(4 Epoch만에 테스트셋 애큐러시 89.58%)

```
Epoch 1/4
40/40 [=====] - 10s 261ms/step - loss: 2.9676 - acc: 0.6748 - val_loss: 1.8946 - val_acc: 0.7969
Epoch 2/4
40/40 [=====] - 9s 219ms/step - loss: 2.0105 - acc: 0.8649 - val_loss: 1.7177 - val_acc: 0.8802
Epoch 3/4
33/40 [=====>.....] - ETA: 1s - loss: 2.0094 - acc: 0.875840/40 [=====] - 9s 219ms/step - loss: 2.0094 - acc: 0.8758
Epoch 4/4
40/40 [=====] - 9s 223ms/step - loss: 2.0601 - acc: 0.8725 - val_loss: 1.7145 - val_acc: 0.8958
```

Accuracy: 89.58%

개선 IDEA - VGG16 Model

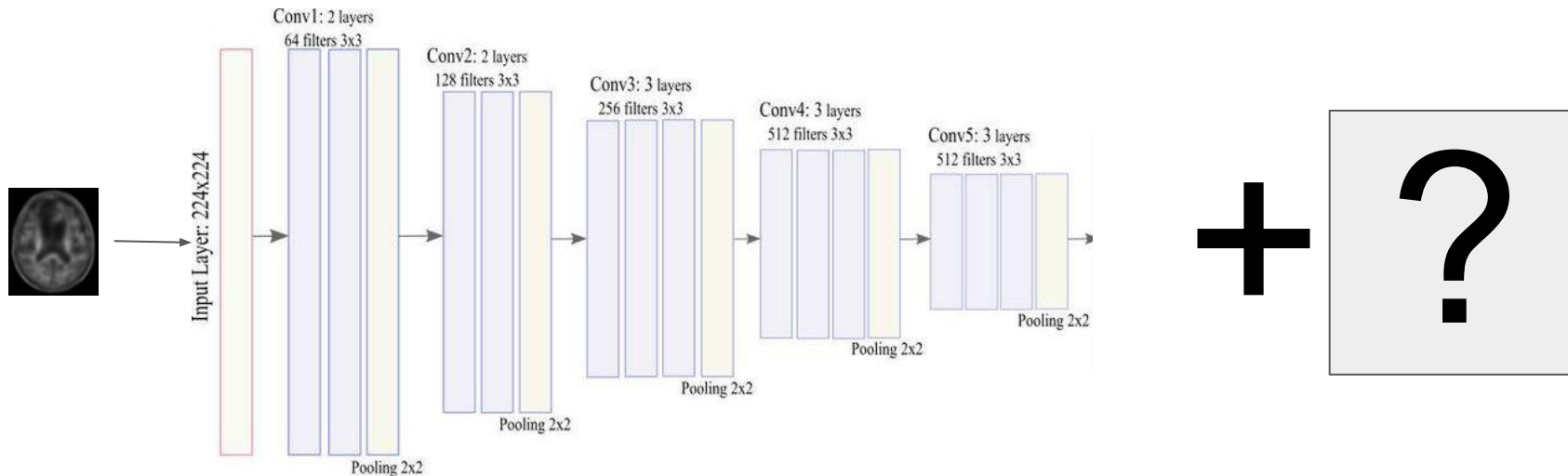
VGG16 모델로 실험!



VGG16 모델의 수리적 포인트는 단순한 패치인 3 x 3 CONVOLUTION MAP을 일렬로 이어붙여서

COMPUTATIONAL COST ↓ NONLINEARITY ↑ 를 노려봄

개선 IDEA - VGG16 Model Transfer Learning



개선 IDEA - Hyperparameter Setting

Learning rate

Layer Numbers

Nodes for each layer

Epoch

+



어떻게
정하지??

개선 IDEA - Hyperparameter Setting

1. 아무거나 넣어봄

이것저것 넣어 보니 learning rate은 $1e-5$, Layer 수는 2개로.. Node 수는 1024에 dropout을 0.5만큼 넣어 주고 적절히 overffting 되지 않게 epoch을 3정도로 주니 정확도가 80%정도 나오네?? 이걸로 해야지...

개선 IDEA - Hyperparameter Setting

2. Grid Search

Try all Candidates

Learning rate = $[1e-7 \sim 1e-2]$

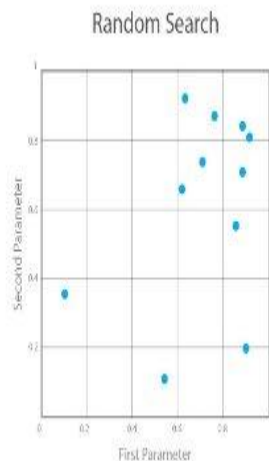
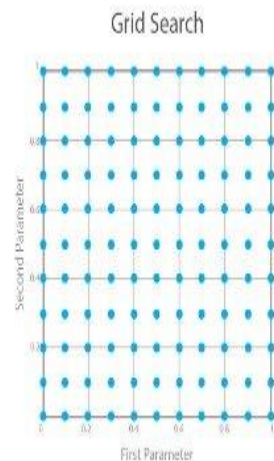
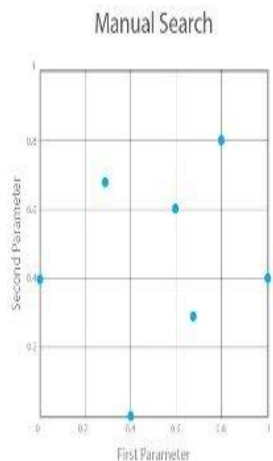
Num of dense layers = $[1 \sim 3]$

Num of nodes for each layer = $[5 \sim 1024]$

Num of epoch = $[1 \sim 20]$

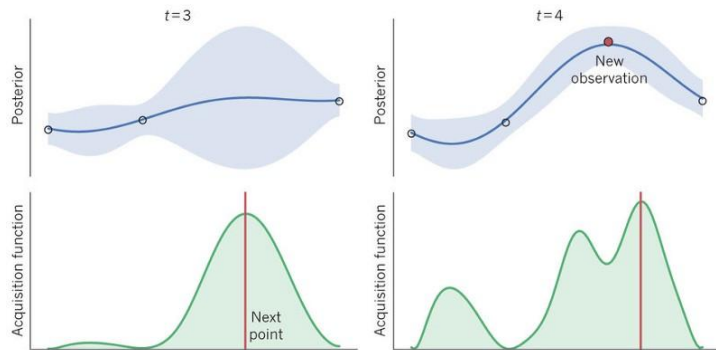
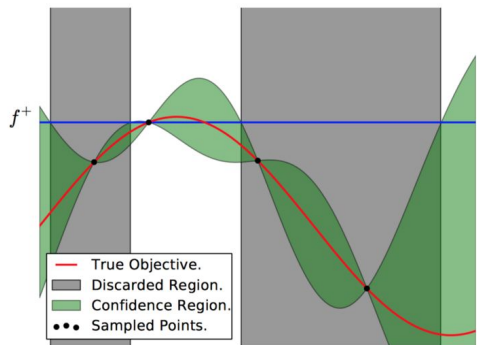
한 모델 학습시키는 데에만 10시간이
걸리는데

언제 다 학습시키지??...



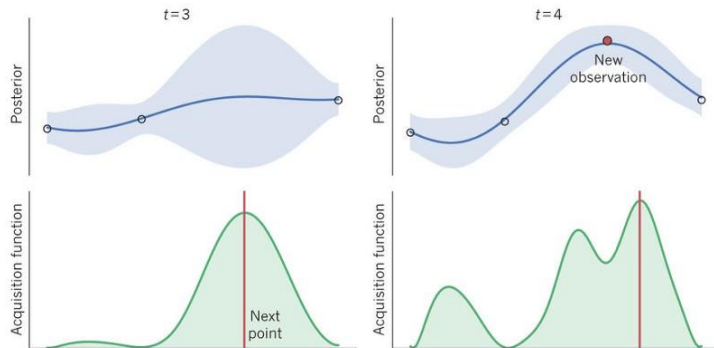
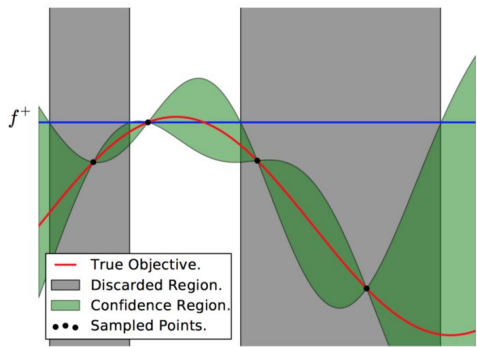
개선 IDEA - Bayesian Optimization

3. Bayesian Optimization



Accuracy Function = $F(\text{learning_rate}, \text{num layers}, \text{num nodes}, \text{num epochs})$

개선 IDEA - Bayesian Optimization



Accuracy Function = $F(\text{learning_rate}, \text{num layers}, \text{num nodes}, \text{num epochs})$

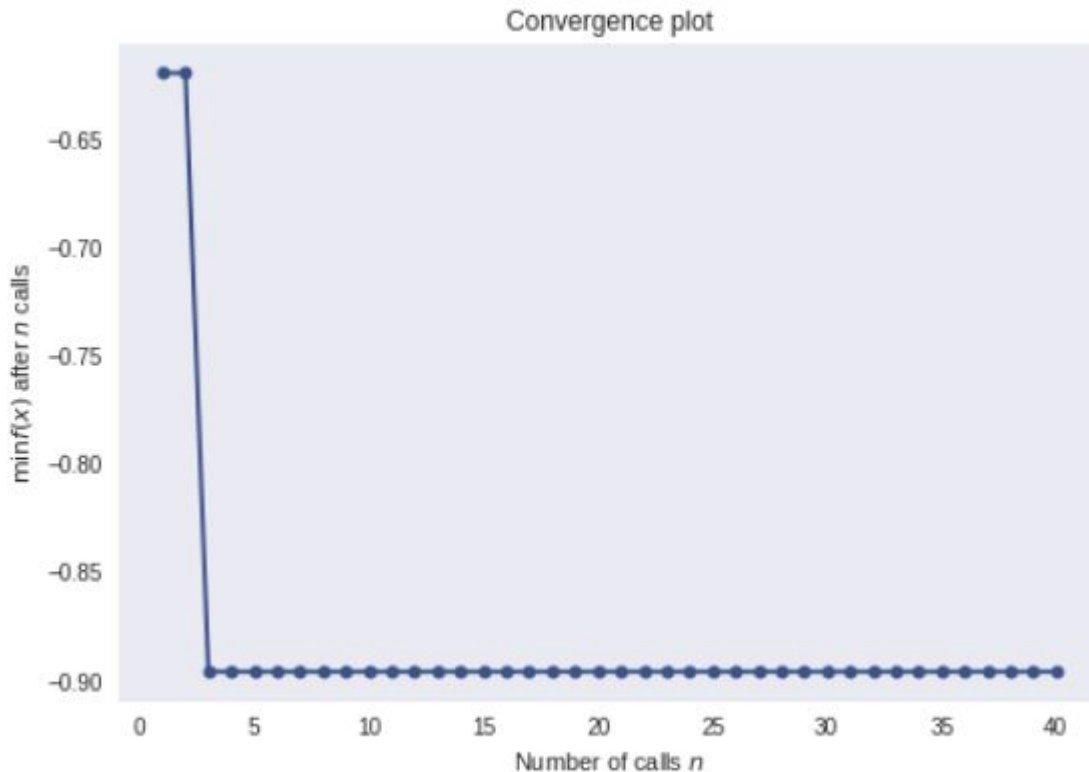
사전 결과값을 바탕으로 사후 확률인 **Accuracy**를 빠르게 찾아냄

개선 IDEA - Bayesian Optimization

다섯 번의 모델
학습만으로도

하이퍼파라미터들을

추정할 수 있음!



개선 IDEA - Bayesian Optimization

```
learning rate: 8.8e-04  
num_dense_layers: 1  
num_dense_nodes: 268  
num_epoch: 4
```

```
Epoch 1/4
```

```
40/40 [=====] - 10s 261ms/step - loss: 2.9676 - acc: 0.6748 - val_loss: 1.8946 - val_acc: 0.7969
```

```
Epoch 2/4
```

```
40/40 [=====] - 9s 219ms/step - loss: 2.0105 - acc: 0.8649 - val_loss: 1.7177 - val_acc: 0.8802
```

```
Epoch 3/4
```

```
33/40 [=====>.....] - ETA: 1s - loss: 2.0094 - acc: 0.875840/40 [=====] - 9s 219ms/step - loss:
```

```
Epoch 4/4
```

```
40/40 [=====] - 9s 223ms/step - loss: 2.0601 - acc: 0.8725 - val_loss: 1.7145 - val_acc: 0.8958
```

```
Accuracy: 89.58%
```


개선 IDEA - Bayesian Optimization

Bayesian Optimization 결과

Learning rate = $8.8e-04$

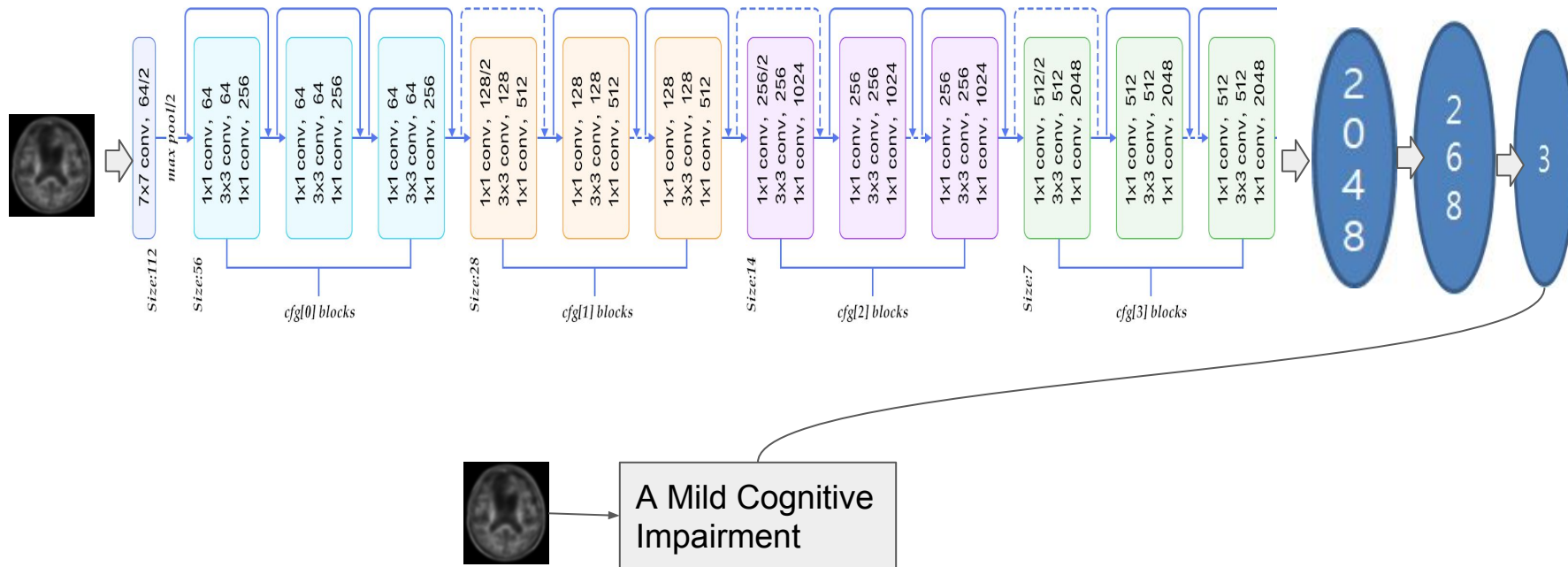
Number of dense_layer = 1

num_dense_nodes = 268

num_epoch = 4 에서

Accuracy : 89.58%

개선 결과

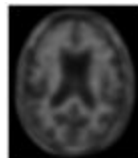


개선 결과

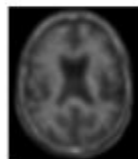
Confusion matrix:

```
[[73  0  0]
 [16  0  4]
 [ 0  0 59]]
```

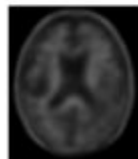
| | precision | recall | f1-score | support |
|-------------|-----------|--------|----------|---------|
| 0 | 0.82 | 1.00 | 0.90 | 73 |
| 1 | 0.00 | 0.00 | 0.00 | 20 |
| 2 | 0.94 | 1.00 | 0.97 | 59 |
| avg / total | 0.76 | 0.87 | 0.81 | 152 |



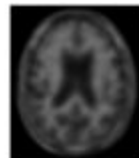
True: 2gr
Pred: 1gr



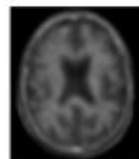
True: 2gr
Pred: 1gr



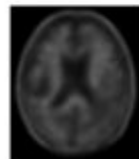
True: 2gr
Pred: 1gr



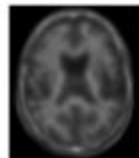
True: 2gr
Pred: 1gr



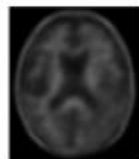
True: 2gr
Pred: 3gr



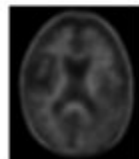
True: 2gr
Pred: 1gr



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Pred: 1gr



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개선 결과

Confusion matrix:

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```

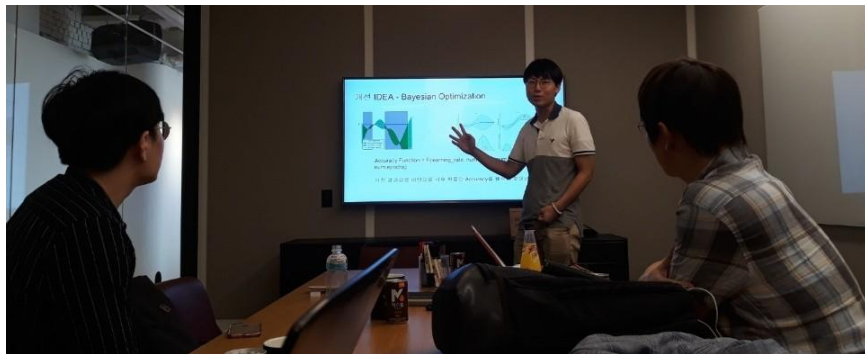
| | precision | recall | f1-score | support |
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| 0 | 0.82 | 1.00 | 0.90 | 73 |
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| 2 | 0.94 | 1.00 | 0.97 | 59 |
| avg / total | 0.76 | 0.87 | 0.81 | 152 |

0단계와 2단계는 우리가 만든
‘알츠넷’(명칭)가 잘 분류해 냄

그런데 1단계는 이 녀석이 정도가 **Mild**
하면 0단계로

정도가 약간 **Severe**하면 2단계로 분류해
버림. -> 사실 사람들도 비슷하지 않나?

1단계 샘플에 대한 심층 분석이 필요함



- 감사합니다 -