# 캡스톤 디자인 '딥메이크 탐지'

## #9. Adversarial training III

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## 지난 캡스톤 회의 내용

● 지난번 adversarial training 의 잘못된 점을 깨달아 다시 실험 진행했음

=> 이번주 진행한 내용

- train: 5000, valid: 1800, test: 1800장으로 실험 진행
- precision, recall 을 넣기로 했으나 문제점 발생

## precision, recall

#### - 사이킷런 패키지 이용

warn prf(average, modifier, msg start, len(result)) warn prf(average, modifier, msg start, len(result))

\_warn\_prf(average, modifier, msg\_start, len(result))

\_warn\_prf(average, modifier, msg\_start, len(result))

\_warn\_prf(average, modifier, msg\_start, len(result))

```
from sklearn.metrics import recall score, precision score
precision score = precision score(target.data.cpu().numpy(), pred.cpu().numpy())
 recall_score = recall_score(target.data.cpu().numpy(), pred.cpu().numpy())
log = 'loss - {:.4f}, acc - {:.3f}, precision - {:.3f}, recall - {:.3f}'.format(epoch loss, epoch acc, precision score, recall score)
Epoch 3/3
                                  | 0/313 [00:00<?, ?it/s]/usr/local/lib/python3.7/dist-packages/torch/utils/data/dataloader.py:481
Train: 0%
  cpuset checked))
Train: 100%|
                                   313/313 [14:07<00:00, 2.71s/it, loss - 0.0049, acc - 0.998, precision - 1.000, recall - 1.000]
                 | 1/113 [00:10<19:49, 10.62s/it, loss - 0.0000, acc - 1.000, precision - 0.000, recall - 0.000]/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classificate
  warn prf(average, modifier, msg_start, len(result))
 usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:1318: UndefinedMetricWarning: Recall is ill-defined and being set to 0.0 due to no true samples. Use `zero div
  _warn_prf(average, modifier, msg_start, len(result))
                | 2/113 [00:11<09:18, 5.03s/it, loss - 0.0000, acc - 1.000, precision - 0.000, recall - 0.000]/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classificati
 usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:1318: UndefinedMetricWarning: Recall is ill-defined and being set to 0.0 due to no true samples. Use `zero div
                 | 3/113 [00:12<05:46, 3.15s/it, loss - 0.0000, acc - 1.000, precision - 0.000, recall - 0.000]/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classificat
  _warn_prf(average, modifier, msg_start, len(result))
                                                                                                                                                      - train은 제대로 나오지만.
 /usr/local/lib/python3.7/dist-packages/sklearn/metrics/ classification.py:1318: UndefinedMetricWarning: Recall is ill-defined and being set to 0.0 due to no true samples. Use `zero divi
  warn prf(average, modifier, msg start, len(result))
                | 4/113 [00:13<04:08. 2.28s/it.loss - 0.0000.acc - 1.000.precision - 0.000, recall - 0.000]/usr/local/lib/python3.7/dist-packages/sklearn/metrics/ classification
 warn prf(average, modifier, msg start, len(result))
                                                                                                                                                     - valid의 경우 한장한장 측정한
 usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification.py:1318: UndefinedMetricWarning: Recall is ill-defined and being set to 0.0 due to no true samples. Use `zero_divi
 _warn_prf(average, modifier, msg_start, len(result))
 /alid: 4%|
                | 5/113 [00:14<03:10, 1.76s/it, loss - 0.0000, acc - 1.000, precision - 0.000, recall - 0.000]/usr/local/lib/python3.7/dist-packages/sklearn/metrics/_classification
                                                                                                                                                      결과로 나옴
```

/usr/local/lib/python3.7/dist-packages/sklearn/metrics/\_classification.py:1318: UndefinedMetricWarning: Recall is ill-defined and being set to 0.0 due to no true samples. Use `zero\_divi

/usr/local/lib/python3.7/dist-packages/sklearn/metrics/ classification.py:1318: UndefinedMetricWarning: Recall is ill-defined and being set to 0.0 due to no true samples. Use `zero div

/usr/local/lib/python3.7/dist-packages/sklearn/metrics/\_classification.py:1318: UndefinedMetricWarning: Recall is ill-defined and being set to 0.0 due to no true samples. Use `zero\_div

| 6/113 [00:15<02:33. 1.44s/it.loss - 0.0000.acc - 1.000.precision - 0.000].recall - 0.000]/usr/local/lib/python3.7/dist-packages/sklearn/metrics/ classification

| 7/113 [00:16<02:18, 1.30s/it, loss - 0.0000, acc - 1.000, precision - 0.000]/usr/local/lib/python3.7/dist-packages/sklearn/metrics/ classificat

| 8/113 [00:17<02:03, 1.18s/it, loss - 0.0000, acc - 1.000, precision - 0.000, recall - 0.000]/usr/local/lib/python3.7/dist-packages/sklearn/metrics/\_classificat

| 9/113 [00:18<01:52. 1.08s/it.loss - 0.0000. acc - 1.000. precision - 0.000. recall - 0.000]/usr/local/lib/python3.7/dist-packages/sklearn/metrics/ classificat

## precision, recall

epoch\_loss = running\_loss / float(n)
epoch\_acc = running\_corrects / float(n)
precision = acc\_num / predict\_num
recall = acc\_num / target\_num

recall = (recall.numpy()[0] \* 100).round(3)
precision = (precision.numpy()[0] \* 100).round(3)

```
lef validate(test loader, model, criterion):
   n = 0
   running loss = 0.0
   running corrects = 0
   correct = 0
   classnum = 2
   target_num = torch.zeros((1, classnum))
   predict num = torch.zeros((1, classnum))
   acc_num = torch.zeros((1, classnum))
with tgdm.tgdm(valid loader, total=len(valid loader), desc="Valid", file=sys.stdout) as iterator:
   for images, target in iterator:
       if args.gpu is not None:
           images = images.cuda(args.gpu, non_blocking=True)
                                                            Epoch 1/3
           target = target.cuda(args.gpu, non_blocking=True)
                                                            Train: 0%|
                                                                                  | 0/313 [00:00<?, ?it/s]/usr/local/lib/python3.7/dist-packages/torch/utils/data/dataloader.py:481: UserWarning: This DataL
                                                              cpuset checked))
       with torch.no_grad():
                                                            Train: 100%|
                                                                                   313/313 [05:04<00:00, 1.03it/s, ('loss - 0.0122, acc - 0.997', 'recall', '99.72 99.6', 'precision', '99.6 99.72')]
           output = model(images)
                                                            Valid: 100%|
                                                                                   113/113 [01:26<00:00, 1.31it/s, ('loss - 0.0049, acc - 0.999', 'recall', '100.0 99.778', 'precision', '99.778 100.0')]
                                                            Epoch 2/3
                                                            Train: 100%|
                                                                                   313/313 [05:00<00:00, 1.04it/s, ('loss - 0.0095, acc - 0.997', 'recall', '99.66 99.78', 'precision', '99.78 99.66')]
       loss = criterion(output, target)
                                                            Valid: 100%
                                                                                   113/113 [00:52<00:00, 2.17it/s, ('loss - 0.0102, acc - 0.998', 'recall', '100.0 99.611', 'precision', '99.613 100.0')]
       _, pred = torch.max(output.data, 1)
                                                            Epoch 3/3
                                                            Train: 100%|
                                                                                   313/313 [05:00<00:00, 1.04it/s, ('loss - 0.0006, acc - 1.000', 'recall', '99.98 99.98', 'precision', '99.98 99.98')]
       n += images.size(0)
                                                                                   113/113 [00:52<00:00, 2.15it/s, ('loss - 0.0030, acc - 0.999', 'recall', '99.833 100.0', 'precision', '100.0 99.834')]
                                                            Valid: 100%|
       running loss += loss.item() * images.size(0)
       running corrects += torch.sum(pred == target.data)
                                                                                                                       precision, recall 결과에 숫자 2개가 나옴
       correct += pred.eq(target.data).cpu().sum()
       pre mask = torch.zeros(output.size()).scatter (1, pred.cpu().view(-1, 1), 1.)
       predict_num += pre_mask.sum(0)
       tar_mask = torch.zeros(output.size()).scatter_(1, target.data.cpu().view(-1, 1), 1.)
       target num += tar mask.sum(0)
       acc mask = pre mask * tar mask
       acc num += acc mask.sum(0)
```

## Gaussian noise test

생성한 노이즈 데이터셋을 xception 모델로 성능 측정

strong	loss : 0.3368, acc: 0.787
medium	loss : 0.8767, acc : 0.789
weak	loss : 4.2529, acc : 0.751

<noise를 추가한 real데이터셋만으로 추론한 결과><noise를 추가한 fake데이터셋만으로 추론한 결과>strong - acc 0.957strong - acc 0.219medium - acc 1.00medium - acc 0.143weak - acc 1.00weak - acc 0.004

# Salt and pepper noise test

```
1 print('-' * 50)
                 2 acc = validate(valid loader, model, criterion)
strong
                Valid: 100%
                                     || 1800/1800 [08:44<00:00, 3.43it/s, loss - 4.1654, acc - 0.580]
                 1 print('-' * 50)
                 2 acc = validate(valid_loader, model, criterion)
mediu
   m
                                      1800/1800 [08:10<00:00, 3.67it/s, loss - 5.5271, acc - 0.531]
                Valid: 100%
                 1 print('-' * 50)
                 2 acc = validate(valid loader, model, criterion)
 weak
                Valid: 100%
                                     | 1800/1800 [28:00<00:00, 1.07it/s, loss - 1.8494, acc - 0.657]
```

```
<noise를 추가한 real데이터셋만으로 추론한 결과>
strong - acc 0.999
medium - acc 1.00
weak - acc 1.00
```

<noise를 추가한 fake데이터셋만으로 추론한 결과>
strong - acc 0.156
medium - acc 0.060
weak - acc 0.321

# **Sharpening noise test**

```
<noise를 추가한 real데이터셋만으로 추론한 결과>
    strong - acc 1.00
    medium - acc 1.00
    weak - acc 1.00
```

```
<noise를 추가한 fake데이터셋만으로 추론한 결과>
     strong - acc 0.07
     medium - acc 0.041
     weak - acc 0.09
```

## Gaussian model adversarial train

strong

medium

weak

```
Epoch 1/3

Train: 0% | | 0/313 [00:00<?, ?it/s]/usr/local/lib/python3.7/dist-packages/torcpuset_checked))

Train: 100% | | 313/313 [12:03<00:00, 2.31s/it, loss - 0.1652, acc - 0.933]

Yalid: 100% | | 113/113 [01:35<00:00, 1.18it/s, loss - 0.6136, acc - 0.829]

Epoch 2/3

Train: 100% | | 313/313 [11:25<00:00, 2.19s/it, loss - 0.0292, acc - 0.991]

Yalid: 100% | | 113/113 [01:28<00:00, 1.28it/s, loss - 0.1913, acc - 0.334]

Epoch 3/3

Train: 100% | | 313/313 [11:26<00:00, 2.19s/it, loss - 0.0207, acc - 0.993]

Yalid: 100% | | 113/113 [01:27<00:00, 1.28it/s, loss - 0.3253, acc - 0.898]
```

```
Epoch 1/3
Train: 0% | 0/313 [00:00<7, ?it/s]/usr/local/lib/python3.7/dist-packages/torch cpuset_checked))
Train: 100% | 313/313 [05:27<00:00, 1.05s/it, loss - 0.0126, acc - 0.996]
Valid: 100% | 113/113 [01:50<00:00, 1.02it/s, loss - 0.0544, acc - 0.978]
Epoch 2/3
Train: 100% | 313/313 [04:55<00:00, 1.06it/s, loss - 0.0043, acc - 0.998]
Valid: 100% | 113/113 [00:51<00:00, 2.18it/s, loss - 0.1482, acc - 0.958]
Epoch 3/3
Train: 100% | 313/313 [05:01<00:00, 1.04it/s, loss - 0.0078, acc - 0.998]
Valid: 100% | 113/113 [00:54<00:00, 2.09it/s, loss - 0.0474, acc - 0.983]
```

→ 최고 성능

# Salt and pepper adversarial train

## strong

### medium

#### weak

```
Epoch 1/3
                      | 0/313 [00:00<?, ?it/s]/usr/local/lib/python3.7/dist-packages/
Train: 0%1
  cpuset checked))
Train: 100%|
                       313/313 [14:51<00:00, 2.85s/it, loss - 0.0154, acc - 0.995]
Valid: 100%
                       113/113 [01:47<00:00, 1.05it/s, loss - 0.0020, acc - 0.999]
Epoch 2/3
                       313/313 [14:13<00:00, 2.73s/it, loss - 0.0050, acc - 0.998]
Train: 100%|
                       113/113 [01:39<00:00. 1.13it/s, loss - 0.0111, acc - 0.997]
Valid: 100%|
Epoch 3/3
Train: 100%
                       313/313 [14:11<00:00, 2.72s/it, loss - 0.0049, acc - 0.998]
                       113/113 [01:39<00:00, 1.13it/s, loss - 0.0130, acc - 0.995]
Valid: 100%
```

```
Epoch 1/3
Train: 0%|
                      | 0/313 [00:00<?, ?it/s]/usr/local/lib/python3.7/dist-packages/
 cpuset checked))
                       313/313 [11:22<00:00, 2.18s/it, loss - 0.0106, acc - 0.997]
Train: 100%|
                       113/113 [01:45<00:00, 1.07it/s, loss - 0.0028, acc - 0.999]
Valid: 100%1
Epoch 2/3
                       313/313 [10:41<00:00, 2.05s/it, loss - 0.0042, acc - 0.999]
Train: 100%
Valid: 100%
                       113/113 [01:20<00:00, 1.40it/s, loss - 0.0023, acc - 1.000]
Epoch 3/3
                       313/313 [10:41<00:00, 2.05s/it, loss - 0.0047, acc - 0.998]
Train: 100%
                       113/113 [01:21<00:00. 1.38it/s. loss - 0.7852. acc - 0.845]
Valid: 100%
```

```
Epoch 1/3
                      | 0/313 [00:00<?, ?it/s]/usr/local/lib/python3.7/dist-packages/
Train: 0%|
 cpuset checked))
                       313/313 [15:09<00:00, 2.91s/it, loss - 0.0074, acc - 0.998]
Train: 100%|
Valid: 100%|
                       113/113 [01:53<00:00, 1.01s/it, loss - 0.0018, acc - 1.000]
Epoch 2/3
                       313/313 [14:31<00:00, 2.79s/it, loss - 0.0002, acc - 1.000]
Train: 100%|
Valid: 100%|
                       113/113 [01:41<00:00, 1.12it/s, loss - 0.0011, acc - 1.000]
Epoch 3/3
Train: 100%|
                       313/313 [14:29<00:00, 2.78s/it, loss - 0.0045, acc - 0.998]
Valid: 100%
                       113/113 [01:40<00:00, 1.13it/s, loss - 0.0061, acc - 0.998]
```

# Sharpening adversarial train

## strong

#### medium

#### weak

```
Epoch 1/3
 Train: 100%|
                        313/313 [06:16<00:00, 1.20s/it, loss - 0.0149, acc - 0.995]
 Valid: 100%|
                        113/113 [01:03<00:00, 1.79it/s, loss - 0.0288, acc - 0.989]
 Epoch 2/3
Train: 100%
                        313/313 [06:18<00:00, 1.21s/it, loss - 0.0068, acc - 0.998]
 Valid: 100%
                        113/113 [00:52<00:00, 2.13it/s, loss - 2.2591, acc - 0.828]
Epoch 3/3
Train: 100%
                        313/313 [06:34<00:00, 1.26s/it, loss - 0.0068, acc - 0.998]
 Valid: 100%|
                        113/113 [01:00<00:00, 1.88it/s, loss - 0.0100, acc - 0.997]
Epoch 1/3
Train: 100%
                        313/313 [03:57<00:00, 1.32it/s, loss - 0.0185, acc - 0.993]
                        113/113 [00:35<00:00, 3.20it/s, loss - 0.1684, acc - 0.938]
Valid: 100%
Epoch 2/3
Train: 100%
                        313/313 [03:41<00:00, 1.42it/s, loss - 0.0020, acc - 0.999]
Valid: 100%
                        113/113 [00:34<00:00, 3.26it/s, loss - 0.0066, acc - 0.998]
Epoch 3/3
                        313/313 [03:40<00:00, 1.42it/s, loss - 0.0069, ac
                                                                            - 0.9981
Train: 100%
Valid: 100%
                        113/113 [00:34<00:00, 3.31it/s, loss - 0.0057, acr - 0.999]
```

```
Epoch 1/3

Train: 100% | 313/313 [04:15<00:00, 1.22it/s, loss - 0.0144, acc - 0.996]

Valid: 100% | 113/113 [00:43<00:00, 2.62it/s, loss - 0.0021, acc - 1.000]

Epoch 2/3

Train: 100% | 313/313 [04:03<00:00, 1.28it/s, loss - 0.0001, acc - 1.000]

Valid: 100% | 113/113 [00:35<00:00, 3.17it/s, loss - 0.0090, acc - 0.996]

Epoch 3/3

Train: 100% | 313/313 [03:33<00:00, 1.46it/s, loss - 0.0093, acc - 0.997]

Valid: 100% | 113/113 [00:36<00:00, 3.09it/s, loss - 0.0042, acc - 0.999]
```

	sharpening (strong)	sharpening (medium)	sharpening (weak)	salt & pepper noise (strong)	salt & pepper noise (medium)	salt & pepper noise (weak)
gaussian	loss -	loss -	loss -	loss -	loss -	loss -
noise	23.4648	10.7281	6.8323	0.8668,	0.2872,	0.1553,
(strong)	acc - 0.5	acc - 0.597	acc - 0.690	acc - 0.804	acc - <b>0.921</b>	acc - 0.969
gaussian	loss -	loss -	loss -	loss -	loss -	loss -
noise	25.8483	15.3691	9.3819	1.3565	0.8106	0.2668
(medium)	acc - 0.5	acc - 0.533	acc - 0.621	acc - 0.682	acc - 0.812	acc - 0.939
gaussian	loss -	loss -	loss -	loss -	loss -	loss -
noise	24.210	20.8601	12.0693	22.7093	21.6682	10.1570
(weak)	acc - 0.5	acc - 0.5	acc - 0.527	acc - 0.5	acc - 0.5	acc - 0.503

	sharpening (strong)	sharpening (medium)	sharpening (weak)	gaussian noise (strong)	gaussian noise (medium)	gaussian noise (weak)
salt & pepper noise (strong)	loss - 4.7843 acc - 0.566	loss - 3.6610 acc - 0.699	loss - 2.6289 acc - 0.800	loss - 2.7121 acc - 0.599	loss - 1.9141 acc - 0.647	loss - 0.0554 acc - 0.988
salt & pepper noise (medium)	loss - 33.099 acc - 0.500	loss - 29.290 acc - 0.500	loss - 21.780 acc - 0.500	loss - 55.707 acc - 0.500	loss - 45.779 acc - 0.500	loss - 12.526 acc - 0.501
salt & pepper noise (weak)	loss - 13.543 acc - 0.506	loss - 9.6646 acc - 0.518	loss - 6.2188 <u>acc - 0.575</u>	loss - 13.347 acc - 0.500	loss - 12.629 acc - 0.500	loss - 3.4392 acc - 0.551

<sup>-</sup> salt & pepper noise strong 모델의 sharpening, gaussian noise의 weak에 대한 성능이 다른것에 비해 높음

- salt & pepper noise strong 모델로 측정한 성능이 전반적으로 높음

	salt & pepper noise (strong)	salt & pepper noise (medium)	salt & pepper noise (weak)	gaussian noise (strong)	gaussian noise (medium)	gaussian noise (weak)
sharpening	loss - 0.41	loss - 0.48	loss - 0.214	loss - 1.057	loss - 0.424	loss - 0.176
(strong)	<u>acc - 0.865</u>	<u>acc - 0.854</u>	acc - 0.928	acc - 0.572	acc - 0.768	acc - 0.926
sharpening	loss - 2.67	loss - 1.99	loss - 0.3691	loss - 0.594	loss - 0.562	loss - 0.391
(medium)	<u>acc - 0.546</u>	acc - 0.594	acc - 0.858	acc - 0.695	acc - 0.701	acc - 0.814
sharpening	loss - 0.51	loss - 0.49	loss - 0.114	loss - 0.652	loss - 0.63	loss - 0.45
(weak)	<u>acc - 0.803</u>	acc - 0.824	acc - 0.952	acc - 0.538	<u>acc - 0.567</u>	<u>acc - 0.85</u>