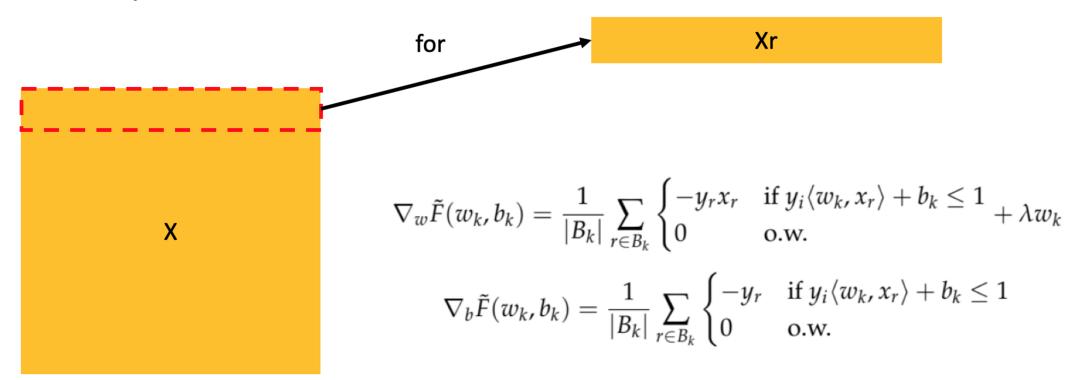
Artificial Intelligence PBL3

01. SVM using GPU

>> Using "for loop" (numpy)

_SGD(X, y, W, b, batch_idx)



>> Not using "for loop" (cupy)

02

03

_SGD(X, y, W, b, batch_idx)

04

X

$$\nabla_w \tilde{F}(w_k, b_k) = \frac{1}{|B_k|} \begin{cases} -yx & \text{if } y * (W^T \cdot X + b) \le 1 \\ 0 & \text{o.w.} \end{cases} + \lambda w_k$$

$$\nabla_b \tilde{F}(w_k, b_k) = \frac{1}{|B_k|} \begin{cases} -y & \text{if } y * (W^T \cdot X + b) \le 1 \\ 0 & \text{o.w.} \end{cases}$$

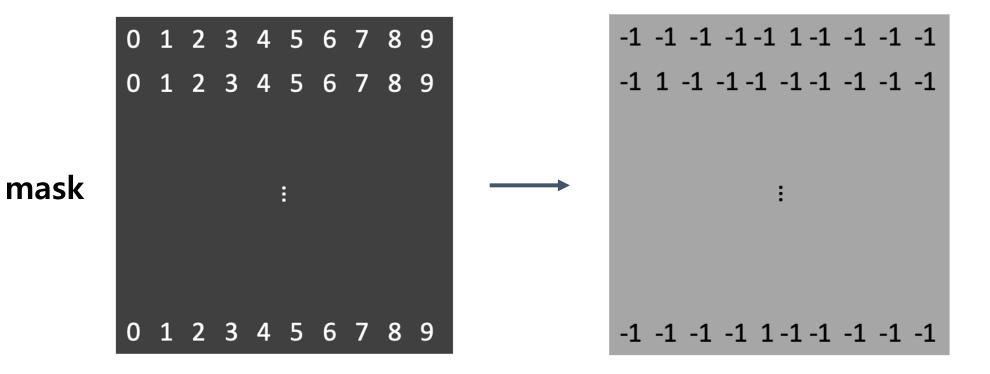
>> Not using "for loop" (cupy)

02

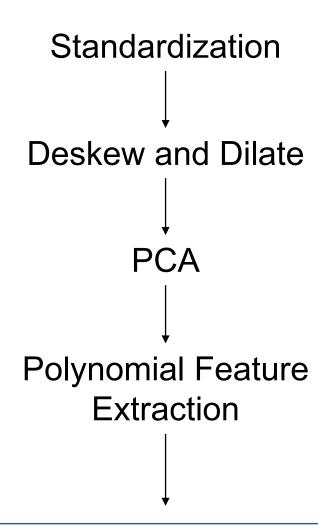
one_hot(y)

У

04



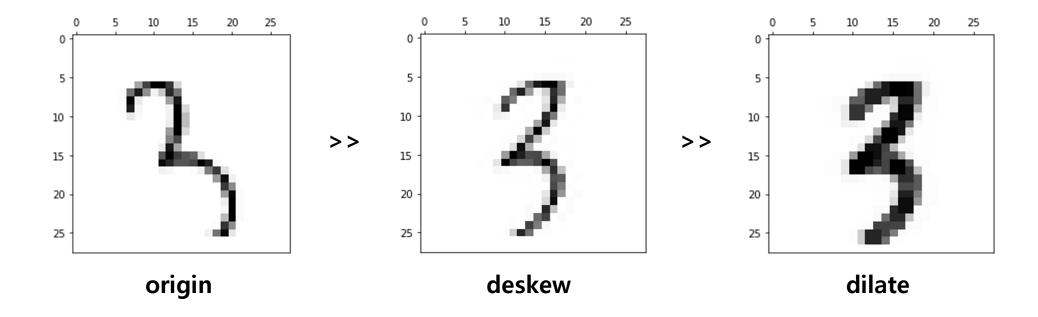
02. Preprocess & Feature Extraction



X_train: (80000, 7924), X_test: (60000, 7924)

>> Preprocessing - Deskew and Dilate

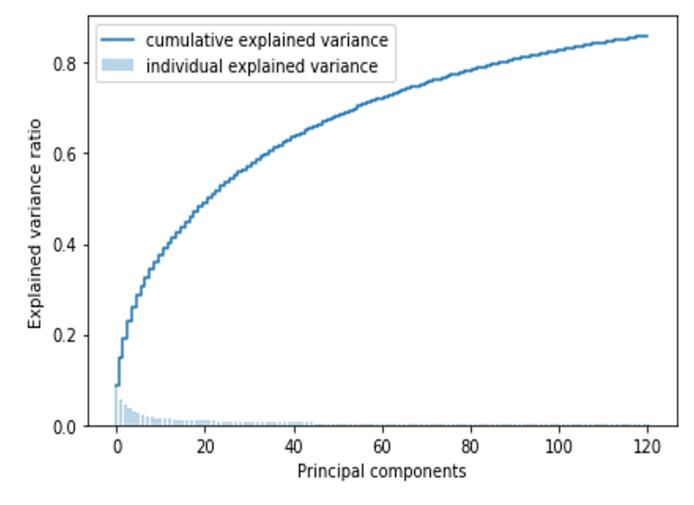
cv2.dilate(img, kernel=(2,2), iterations = 1)



>> Principal Components Analysis

Principal component	Eigenvalue	Cumulative variance	
1	69.7181	0.08926	
2	45.2179	0.1471	
3	35.4805	0.1925	
4	28.8478	0.2295	
5	24.0394	0.2603	
120	1.1345	0.8590	
121	1.1182	0.8604	

$PCA(n_{components} = 0.86)$

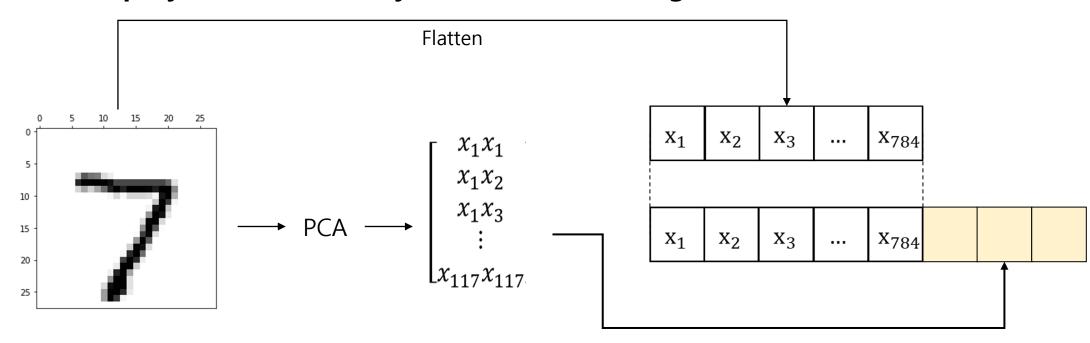


>> Polynomial Feature Extraction

from sklearn.preprocessing import

PolynomialFeatures

poly_features = PolynomialFeatures(degree=2)

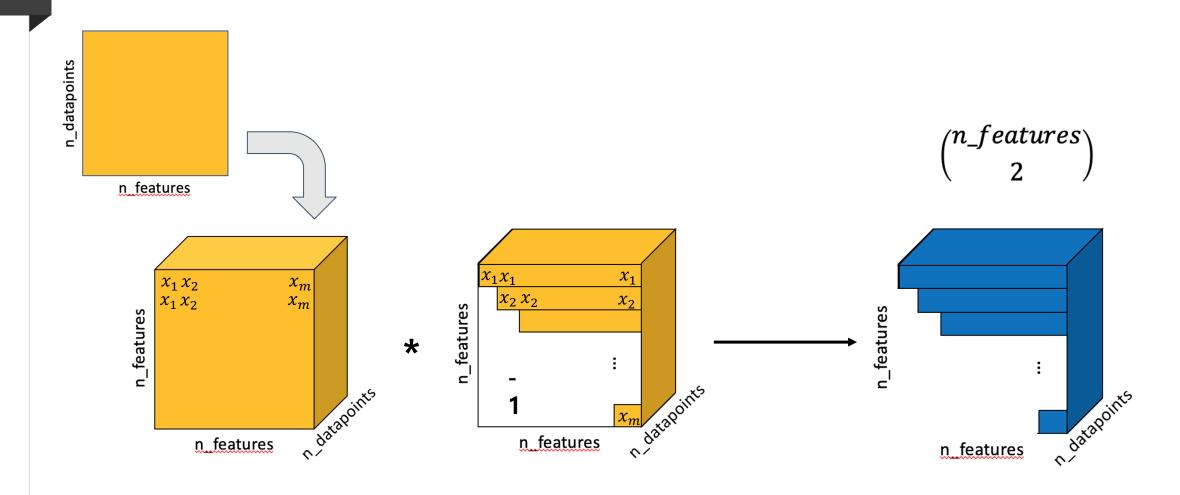


>> Polynomial Feature Extraction - our implementation

Už

03

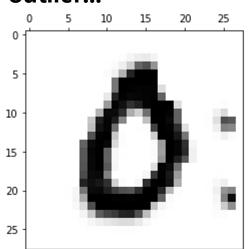
04

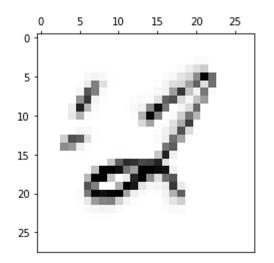


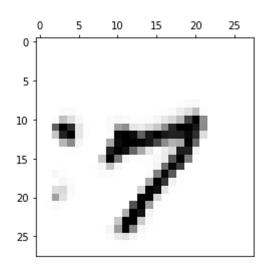
-> Too much computation!!

>> Polynomial Feature Extraction crop image to reduce features and remove outlier

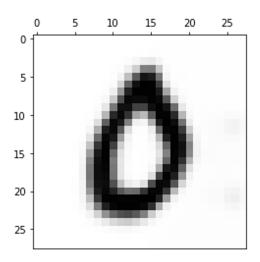


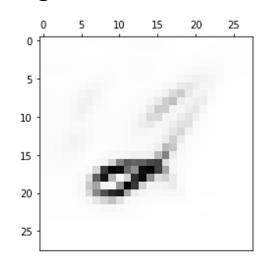


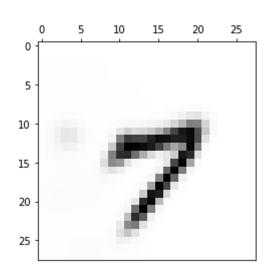




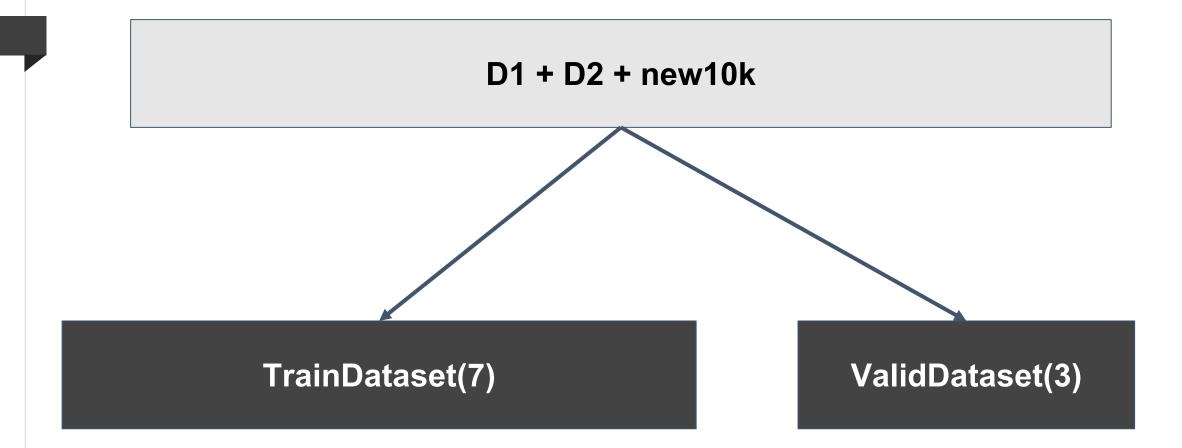
cv2.fastNlMeansDenoising(img, None, 50.0, 7, 21)







03. Hyperparameter Tuning



```
01
```

04

03

```
>> Hyperparameter tuning
```

```
pbl2 - max_iter=200
                                             pbl3 - max_iter=500
```

12min 14s

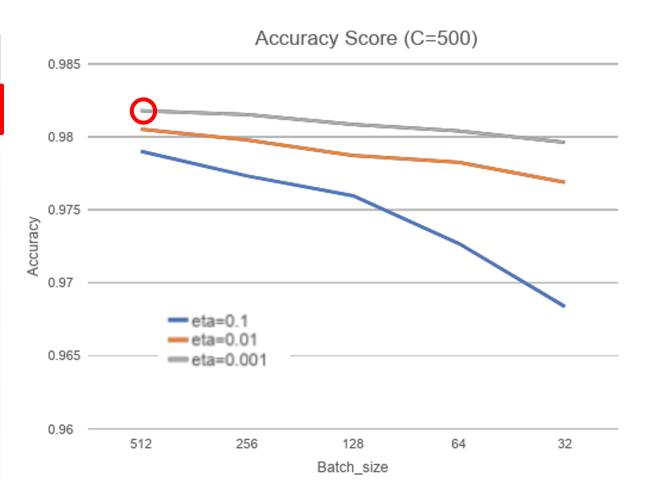
1min 55s

```
param_grid = [{
```

```
'C': [100, 200, 300, 500, 700, 1000]
'eta': [0.1, 0.01, 0.001]
'max_iter' : [500]
'batch_size': [16, 32, 64, 128, 256, 512]
```

>> Hyperparameter tuning

Rank	С	eta	Batch_size	Accuracy
1	500	0.001	512	0.981792
2	500	0.001	256	0.981533
3	1000	0.001	512	0.98148
4	700	0.001	256	0.981375
5	100	0.01	512	0.981317
6	1000	0.001	256	0.98125
7	200	0.01	512	0.980992
:	:	:	:	:



Q & A

Thank you:)