

report from Group 2

- **Kato** and **Sato** report **results and findings** from **Challenge A : Particle Classification**
- **Ninomiya** introduces his motivation to study ML

Challenge A: Particle classifier

- Starting from MNIST code, and adopted small difference of data structure.
- Checked how the performance changes according to change of hyper parameters
- Baseline: Num_filters=16 (and doubled for each MaxPooling), three convolution layers
A: Num_filters = 32
B: Put two Conv2d+LeakyReLU before MaxPool2d
C: MaxPool (4,4) -> (2,2) and 7 convolution layers

Baseline

```
class CNN(torch.nn.Module):
    def __init__(self, num_filters=16):

        super(CNN, self).__init__()
        # feature extractor CNN
        self._feature_extractor = torch.nn.Sequential(
            torch.nn.Conv2d(1, num_filters, 3, padding=1), #in_filters, #out_filters, kernel_size, stride, padding
            torch.nn.LeakyReLU(),
            torch.nn.MaxPool2d(4, 4),
            torch.nn.Conv2d(num_filters, num_filters*2, 3, padding=1),
            torch.nn.LeakyReLU(),
            torch.nn.MaxPool2d(4, 4),
            torch.nn.Conv2d(num_filters*2, num_filters*4, 3, padding=1),
            torch.nn.LeakyReLU(),
            torch.nn.MaxPool2d(12, 12))
        # classifier MLP
```

Model	Accuracy (%)
Baseline	86.9
A	89.9
B	91.6
C	90.5

- All of changes improved the performance.
 - Putting two convolution layers has biggest improvement
- Tried combining A+B+C, but not finished within time..

Challenge A

K. Sato

Based on [IntroNeuralNetwork/Introduction03-MNIST-CNN](#)

I designed ***_feature_extractor*** like ...

_feature_extractor

** (1 for 0-th loop)*

repeat N times

$\text{conv2D}(2^{(i-1)}a_0, 2^i a_0, 3) \rightarrow \text{LeakyLeRU}() \rightarrow \text{MaxPool2d}(b_i, b_i)$

(N+1)-th cycle

$\rightarrow \text{conv2D}(2^{(N-1)}a_0, 2^N a_0, 3) \rightarrow \text{LeakyLeRU}() \rightarrow \text{MaxPool2d}(b_N, b_N)$

remaining pixels

What I changed

- # repetitions **N**
- combinations of **b_i**
- initial # of filters : **a_0**

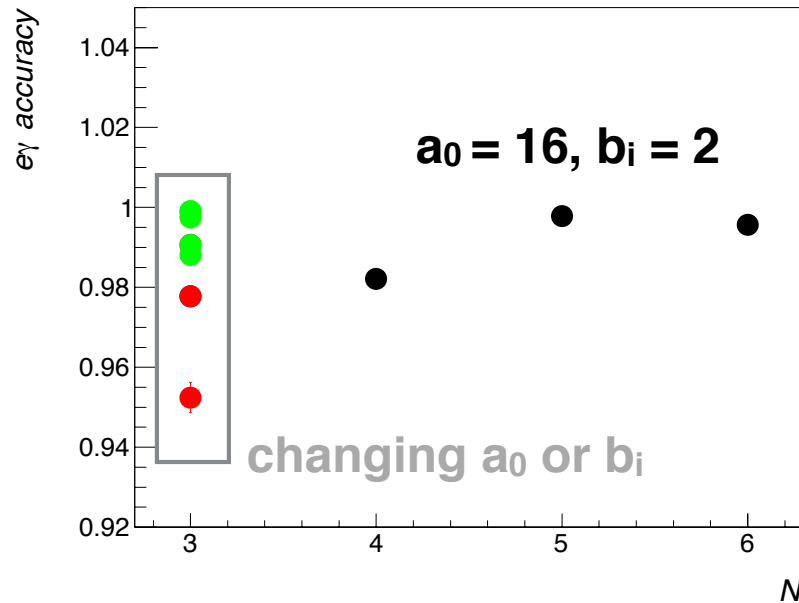
What I checked

- accuracy of (e, γ)
 $(N_{ee} + N_{e\gamma} + N_{\gamma e} + N_{\gamma\gamma}) / (N_e^{\text{true}} + N_\gamma^{\text{true}})$

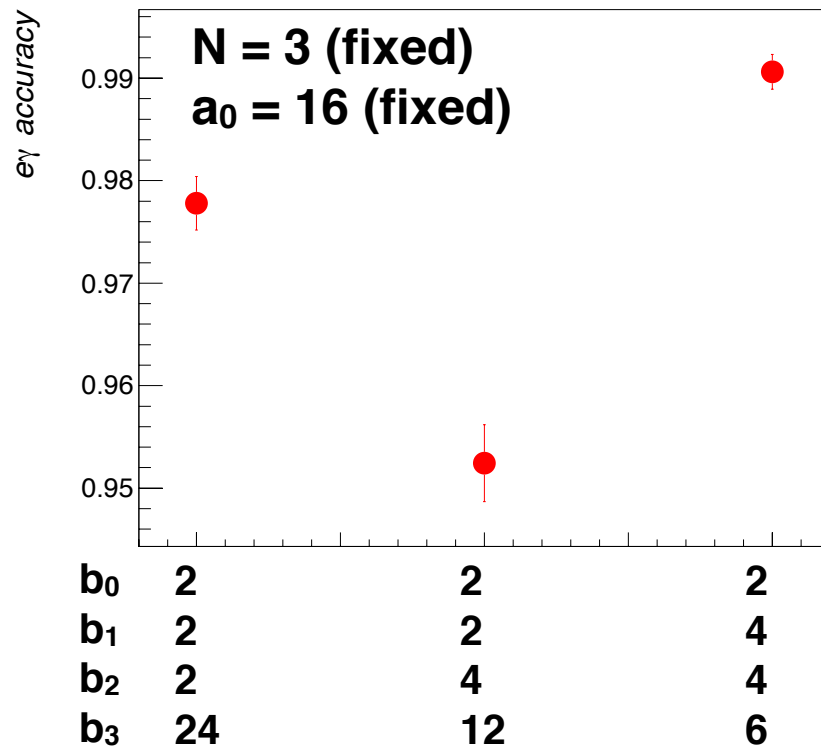
***detector responses for e and γ are quite similar**

result

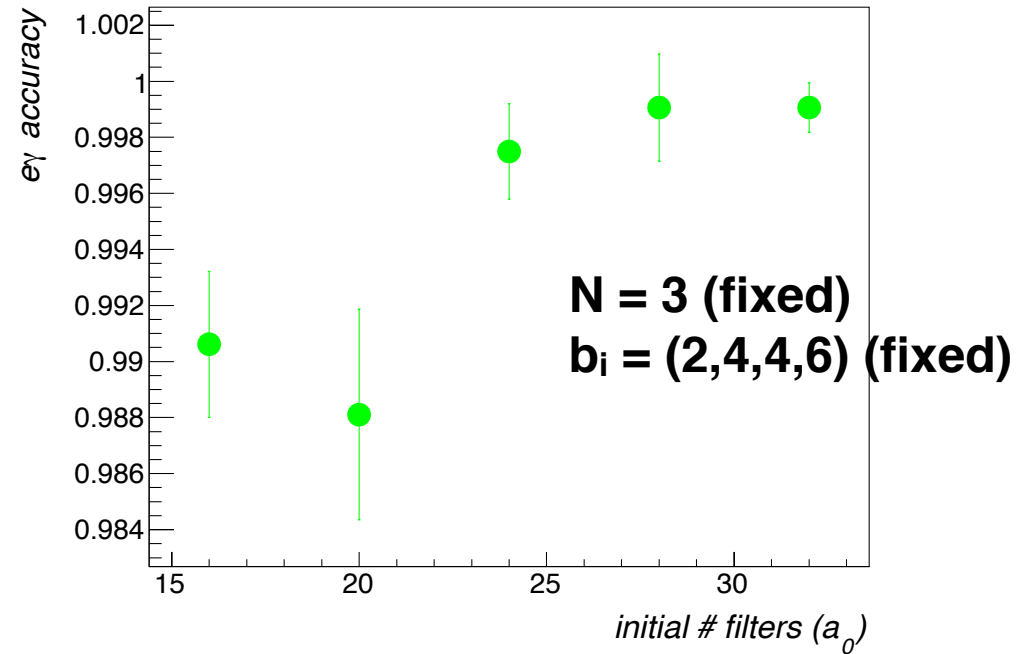
changing # of loop: N



changing b_i in MaxPool2d(b_i, b_i)



changing initial number of filters : a_0



- Larger a_0 (i.e., larger discriminants), better accuracy.
- MaxPool2d(b, b) with larger b reduces accuracy.
 - maybe some information are lost

My B4 experiment

- **Main theme**

Increase the accuracy to reconstructing events from image of water Cherenkov radiation at Super-Kamiokande.

- **Methods**

First step Compare $\nu_\mu/\bar{\nu}_\mu$ identification accuracy of using Maximum likelihood estimation to Neural Network.

Second step Challenging the main theme!

- **Leaning in KMI school**

I learned basis of Neural Network. In particular, I will refer to code for images in MNIST-solution.