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# Collision Avoidance (DNN)

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# Simplify the dimension

- Input features:  $32 \times 32 \times 3$  (3072)  $\rightarrow 32 \times 32 \times 1$  (1024)

# Change the DNN model

- Variable hidden layers
- Batch normalization
- Dropout (same value for every hidden layers)



# Change the DNN model

```
INPUT_SIZE = IMAGE_HEIGHT * IMAGE_WIDTH * IMAGE_CHANNEL
class DNN(nn.Module):
   __slots__ = "__model"
   def __init__(self, input_dim=INPUT_SIZE, output_dim=2, hidden_dims=(128, 64, 32), do_batch_normal=True, dropout=0):
       super(DNN, self).__init__()
       dims_list = (input_dim, *hidden_dims)
       model components = []
        # hidden layers
        for i in range(1, len(dims_list)):
            current_input_dim = dims_list[i-1]
           current_output_dim = dims_list[i]
            model_components.append(nn.Linear(current_input_dim, current_output_dim))
            if do_batch_normal == True:
                model_components.append(nn.BatchNorm1d(current_output_dim))
            model_components.append(nn.ReLU())
            if dropout > 0:
                model components.append(nn.Dropout(dropout))
        # output layer
       output_layer = nn.Linear(dims_list[-1], output_dim)
       model_components.append(output_layer)
       model_components.append(nn.Softmax(dim=1))
       # make DNN model
       self.__model = nn.Sequential(*model_components)
   def forward(self, x):
       return self.__model(x)
```



#### **Train Result**

- **Subject**: Effect of the *dropout* 

- **Hidden Layers**: 1280, 720, 160

- Hyper parameters

Epochs: 30, Learning rate: 0.001, Momentum: 0.9, L2 constant: 1e-5

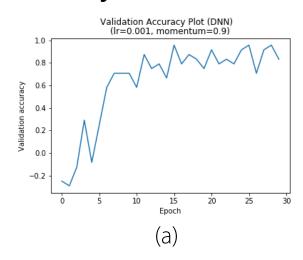
- **Normalization**: Batch normalization only

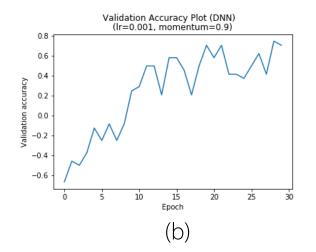
Figure	Dropout	Validation Accuracy	Test Accuracy*
(a)	0.0	0.958	0.875 (21/24)
(b)	0.2	0.750	0.875 (21/24)
(c)	0.4	0.542	0.792 (19/24)



### **Train Result**

- **Subject**: Effect of the *dropout* 





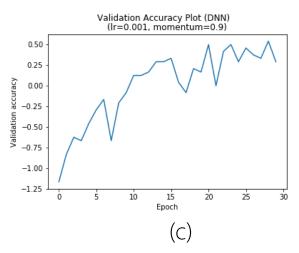


Figure	Dropout	Validation Accuracy	Test Accuracy*
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#### **Train Result**

- **Subject**: Effect of the *dropout* 

Dropout is not useful (In this situation...)



#### **Train Result**

- **Subject**: Effect of the hidden layers

Hyper parameters

Epochs: 30, Learning rate: 0.001, Momentum: 0.9, L2 constant: 1e-5

- **Normalization**: Batch normalization only

- **Dropout**: None (0)

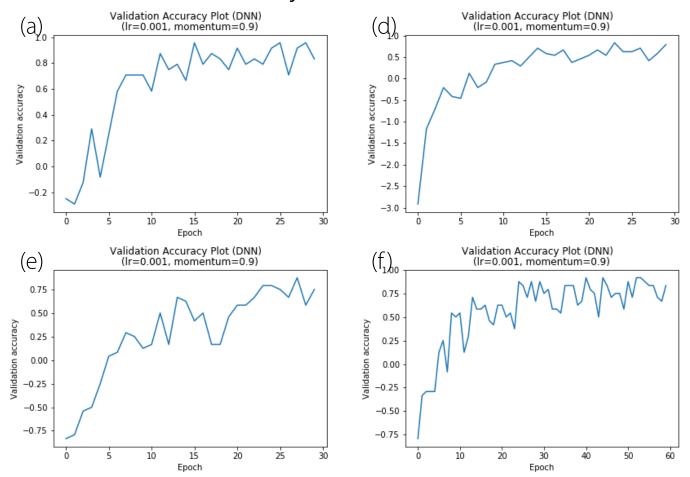
Figure	Hidden layers	Validation Accuracy	Test Accuracy*
(a)	1280, 720, 160	0.958	0.875 (21/24)
(d)	128, 72, 16	0.833	0.917 (22/24)
(e)	128, 64, 16	0.875	0.875 (22/24)
(f)	128, 64, 16	0.917	0.958 (23/24)

Epochs: 60



#### **Train Result**

Subject: Effect of the hidden layers





#### **Train Result**

- **Subject**: Effect of the hidden layers

- 1. 2~3 digits are enough
- 2. We need more than 30 epochs



#### **Train Result**

- **Subject**: Effect of the L2 constant

- **Hidden Layers**: 128, 64, 16

Hyper parameters

Epochs: 30, Learning rate: 0.001, Momentum: 0.9

- Normalization: Batch normalization only

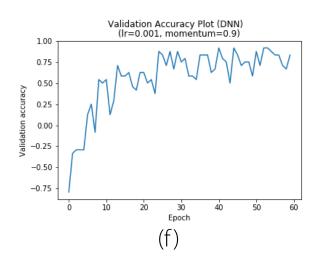
- **Dropout**: None (0)

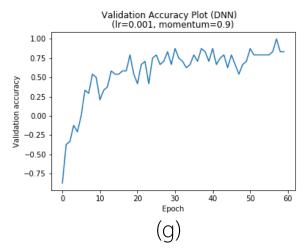
Figure	L2 constant	Validation Accuracy	Test Accuracy*
(f)	1e-5	0.958	0.958 (23/24)
(g)	1e-4	1.000	0.833 (20/24)
(h)	1e-4	0.958	0.958 (23/24)

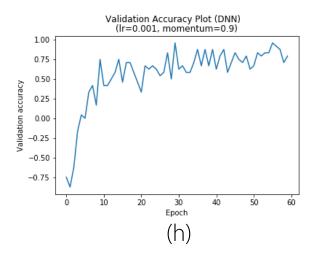


### **Train Result**

- **Subject**: Effect of the L2 constant







No Accuracy benefits

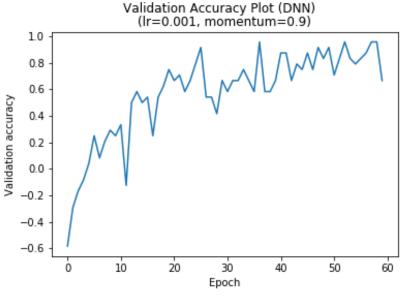


#### **Train Result**

- Add normalization at the transform part

```
total_dataset = ImageFolder(
    DATASET_PATH,
    transforms.Compose([
        transforms.Resize((IMAGE_HEIGHT, IMAGE_WIDTH)),
        transforms.Grayscale(num_output_channels=IMAGE_CHANNEL),
        transforms.ToTensor()
        transforms.Normalize([0.449], [0.226]),
        transforms.Lampda(Tampda Img: torch.flatten(img)) # http
])
)
```

average of the previous 3 elements



Validation accuracy: 0.958 Test accuracy: 0.958 (23/24)



#### **Conclusion**

We can do the image classification by using the custom DNN. However, 95.8% is the maximum accuracy.