

**STAT 453**

**Name:** \_\_\_\_\_

**SS 2025**

**Midterm Exam**

**03/06/2025**

**Time: 2:30 – 4:45 pm am (75 mins) Email:** \_\_\_\_\_ @wisc.edu

**Instructor: Yiqiao Zhong**

**Teaching Assistant: Zhexuan Liu**

**Access Code: ??????-ZHONG**

---

This exam contains 7 pages (including this cover page) and 9 questions.

Total of points is 100.

By submitting this exam, I (the student)

- acknowledge that I am required to follow the academic integrity and conduct policies of UW-Madison.

Grade Table (for teacher use only)

Question	Points	Score
1	10	
2	10	
3	10	
4	10	
5	10	
6	10	
7	10	
8	10	
9	20	
Total:	100	

---

1. (10 points) Which one of the statements is NOT correct about the perceptron learning algorithm?
  - A. Perceptron algorithm is an iterative algorithm.
  - B. In the training loop, the error of each training example takes value  $-1, 0$ , or  $1$ .
  - C. The perceptron algorithm is not guaranteed to converge even for linearly separable data.
  - D. The perceptron algorithm has a history of over 50 years.

---

Solution:

2. (10 points) What is NOT a feature of PyTorch?
  - A. PyTorch and NumPy share many similarities.
  - B. PyTorch has built-in support for automatic differentiation.
  - C. During the forward pass, a dynamic computation graph is created so that the model can implement back-propagation.
  - D. When using PyTorch, we still need to manually work out the backward computation in a network definition.

---

Solution:

3. (10 points) Which one of the statements is NOT correct about the multi-category cross-entropy (CE) loss and the logistic loss?
  - A. The multi-category CE loss and the logistic loss are strongly connected.
  - B. The logistic loss is not a convex loss, which is derived based on the maximum likelihood estimation.
  - C. For classification problems, the CE loss is preferable to the mean-squared error (MSE) loss.
  - D. Since the CE loss is a smooth function, we can use back-propagation to train neural networks with the CE loss.

4. (10 points) Which one is NOT an advantage of mini-batch stochastic gradient descent (SGD)?
- A. Mini-batch SGD is a deterministic algorithm.
  - B. Mini-batch SGD exploits GPU hardware well.
  - C. Mini-batch SGD is computationally efficient since it makes frequent updates to parameters.
  - D. Mini-batch SGD can often escape local minima due to noisy gradients.

---

Solution:

5. (10 points) Which one is NOT a common regularization technique in deep learning?
- A. Dropout.
  - B. Weight decay.
  - C. Data splitting.
  - D. Early stopping.

---

Solution:

6. (10 points) [Dimension calculation] (i) Suppose that we create a linear layer in PyTorch as follows. What is the expected output?

```
import torch

layer1 = torch.nn.Linear(in_features=5, out_features=8)
print(layer1.weight.shape)
print(layer1.bias.shape)
```

---

Solution:

(ii) Next, we create an input tensor and pass it through the linear layer. What is the expected output?

```
data = torch.arange(40).view(-1, 5).float()
out = layer1(data)
print(out.shape)
```

---

Solution:

7. (10 points) [*Translating code to math*] Suppose that in PyTorch we define a neural network class with three linear layers `linear_1`, `linear_2`, `linear_3`, and an output linear layer `linear_out`. All linear layers are created with NO bias terms. We use the symbols  $W_1, W_2, W_3, W_{\text{out}}$  to represent the weight matrices of these linear layers respectively. We also use the symbol  $\sigma$  to represent the ReLU activation function.

The following code gives the forward pass in the neural network class definition. Given an input  $x$ , can you write down the **mathematical expression** to represent the forward pass using the above symbols?

```
def forward(self, x):
    out = self.linear_1(x)
    out = F.relu(out)
    out = self.linear_2(out)
    out = F.relu(out)
    out = F.relu(self.linear_3(x)) + out
    logits = self.linear_out(out)
    return logits
```

---

Solution:

8. (10 points) Suppose that we have a multilayer perceptron (MLP). In one hidden layer, a 200-dimensional feature vector is mapped to a 32-dimensional feature vector. We initialize weights independent from a normal distribution  $N(0, \sigma^2)$ . According to He initialization, what is the value of  $\sigma$ ? *Hint: the formula is  $\sigma = \sqrt{2/\text{width}}$ .*

---

Solution:

9. (20 points) Please explain in words what the following PyTorch code means.

```
(i) torch.manual_seed(42)
(ii) features = features.view(-1, 28*28).to('cuda:0')
(iii) optimizer.zero_grad()
(iv) train_loader = torch.utils.data.DataLoader(dataset=train_dataset,
batch_size=64, shuffle=True)
(v) net = torch.nn.Sequential(
    torch.nn.Linear(28, 30),
    torch.nn.ReLU(),
    torch.nn.Dropout(0.5),
    torch.nn.Linear(30, 10)
)
```

---

Solution:

