ECE 492-45 Homework 5 (Fall 2021) Instructor: Dr. Chau-Wai Wong Material Covered: LSTM, Gradient Descent, Statistical Learning Basics

- Problem 1 (20 points) [Character-level LSTM] In this Colab notebook file, you will use a recurrent neural network with long short-term memory (LSTM) units to predict the next character based on a Shakespeare writing. The trained model auto-generates texts, which can imitate the writing style of Shakespeare. To start text generation, you should pass a starting char(acter), from which you then generate one character at a time. Examine the following items:
 - 1. Test the trained model by passing different starting strings. Also, train with your own dataset, e.g., writing from another author, and show the results.
 - 2. Use one sentence each to explain what the following functions do: random_chunk() and random_training_set().
 - 3. Draw an unrolled block diagram for the following code segment:

```
for c in range(chunk_len - 1):
out_target = target[c].unsqueeze(0).type(torch.LongTensor)
out, hidden, cell = model(inp[c], hidden, cell)
loss += criterion(out, out_target)
```

Problem 2 (20 points) [Level Curves and Gradient Descent]

a) A set of level curves is shown as follows. Use the dot as the starting point, draw a trajectory of gradient descent steps. Annotate each descent step using a line segment with an arrow at the end. Explicitly draw a tangent line at each step, which can assist you to determine the negative gradient direction. You may vary the descent step size.



b) A cost function is given as $J(w_1, w_2) = \exp(-w_1^2 - 10w_2^2)$. Use the command you learned from Problem 2 of HW1 to draw a contour plot/set of level curves. Illustrate by drawing two initial points and their gradient descent trajectories using a fixed step size. One initial point must lead to fast convergence and the other must lead to slow convergence. (You may draw the trajectories on a printout, or you may do a screenshot, paste it into Powerpoint, draw the arrows and tangent lines, and save it as a PDF file.) **Problem 3** (20 points) [Conditional Expectation, Variance Operator]

a) Given the joint PMF for random variables X and Y in the table, compute the following quanti-

Table 1: Joint PMF, $p_{XY}(x, y)$				
$Y \setminus X$	1	2	3	
0	0.3	0.1	0.3	
1	0.1	0.1	0.1	

ties and tabulate your results: $p_X(x)$, $p_Y(y)$, $p_{X|Y}(x|y)$, $p_{Y|X}(y|x)$, $\mathbb{E}[X]$, $\mathbb{E}[Y]$, $\mathbb{E}[X|Y=y]$, $\mathbb{E}[Y|X=x]$. (Intermediate steps must be shown to receive full points.) Explain the difference between $\mathbb{E}[X|Y=y]$ and $\mathbb{E}[X|Y]$.

b) Prove the following formulas:

$$\operatorname{Var}(aX+b) = a^2 \operatorname{Var}(X), \tag{1a}$$

$$\operatorname{Var}(X+Y) = \operatorname{Var}(X) + \operatorname{Var}(Y) + 2 \operatorname{Cov}(X,Y), \tag{1b}$$

$$\operatorname{Var}(X - Y) = \operatorname{Var}(X) + \operatorname{Var}(Y)$$
, when X and Y are uncorrelated. (1c)

$$\operatorname{Var}\left(\sum_{i} a_{i} X_{i}\right) = \sum_{i} a_{i}^{2} \operatorname{Var}(X_{i}), X_{i} \text{'s uncorrelated. Useful for Problem 3e.}$$
(1d)

You may find the following equations useful: i) the shortcut formulas for variance, $\operatorname{Var}(X) = \mathbb{E}[X^2] - (\mathbb{E}X)^2$; and ii) the covariance, $\operatorname{Cov}(X, Y) = \mathbb{E}[XY] - \mathbb{E}[X]\mathbb{E}[Y]$. Answer the following questions:

- Why does b not appear on the right-hand side of (1a)?
- How does the variance of sum of two random variable compare to the sum of the variance of individual variables when the variables are negatively/anti-correlated? Can you give an extreme example?
- Why is it a plus sign rather than a minus sign on the right-hand side of (1c)?
- **Problem 4** (20 points) (20 points, try it after Wed. lecture) [Curse of Dimensionality] Read the first paragraph of the problem statement of *ESLII-2.4*. Note that we may also write $\mathbf{X} = (X_1, X_2, \ldots, X_p)$, where $X_k \sim \mathcal{N}(0, 1)$ for $k = 1, \ldots, p$. Use a programming language of your choice. To get started, set p = 10. Note that in this problem, all vectors are column vectors.
- a) Write a computer program to randomly draw/generate N = 100 vectors from the template random vector **X**, namely, $\{\mathbf{x}^{(i)}, i = 1, ..., N\}$. Note that each vector should contain pnormality distributed random numbers. Plot all vectors as points in a 3-D space consisting of the first, second, the last coordinates.
- b) Calculate the coordinate value of each point after being projected on to a fixed direction specified by $\mathbf{a} = \mathbf{x}_0 / ||\mathbf{x}_0||$, namely, $z^{(i)} = \mathbf{a}^T \mathbf{x}^{(i)}$. Here, \mathbf{x}_0 is an arbitrary nonzero vector of length p, "T" is the transpose operation, and $z^{(i)} \in \mathbb{R}$. What are the sample mean and sample variance of the projected coordinates $\{z^{(i)}, i = 1, \dots, N\}$?

- c) Repeat a) and b) for $p \in [1, 80]$. You may want to use a for loop to achieve this. Optionally, put your code for parts a) and b) into a function to make your code easier to read. Plot the sample variance of the projected coordinates as a function of p.
- d) Calculate the squared distance of each point to the origin, namely, $d_i^2 = ||\mathbf{x}^{(i)}||^2$. What is the sample mean of $\{d_i^2, i = 1, ..., N\}$? Plot the sample mean of the squared distance as a function of p in the same plot of c). Limit the range of y-axis between 0 and 80. For p = 5, inspect the values of any five d_i^2 's. Do the results in b) and c) match with conclusion drawn in the third paragraph of *ESLII-2.4*?
- e) Use the formulas from Problem 1b, prove that $\operatorname{Var}(Z) = 1$ where $Z = \mathbf{a}^T \mathbf{X}$, and $\mathbb{E}[D^2] = p$ where $D = ||\mathbf{X}||$. Are the theoretical results in this part consistent with the simulated results obtained in c) and d)?

Problem 5 (20 points) [Auto Data Analysis] Complete ISLR-2.4.9.

For Python users, please follow the text book's instructions while referring to the "equivalence" Python code, where you may find the sample code and the comments useful.