Your latest: 95% • Your highest: 95% • To pass you need at least 80%. We keep your highest score.

| 1. | . Which of the following are true about hyperparameter search? | 1/1 point |
|----|---|-----------|
| | ○ When using random values for the hyperparameters they must be always uniformly distributed. | |
| | When sampling from a grid, the number of values for each hyperparameter is larger than when using random values. | |
| | Choosing random values for the hyperparameters is convenient since we might not know in advance which hyperparameters are more important for the problem at hand. | |
| | Choosing values in a grid for the hyperparameters is better when the number of hyperparameters to tune is high since it provides a more ordered way to search. | |
| | Correct Correct. Different problems might be more sensitive to different hyperparameters. | |
| 2. | If it is only possible to tune two parameters from the following due to limited computational resources. Which two would you choose? | |
| | $oxedsymbol{\square}$ The eta parameter of the momentum in gradient descent. | |
| | $ ightharpoons$ α | |
| | Correct Correct. This might be the hyperparameter that most impacts the results of a model. | |
| | $oxed{\ }$ ϵ in Adam. | |
| | $leftilde{oldsymbol{oldsymbol{eta}}}eta_1,eta_2$ in Adam. | |
| | $ig\otimes$ This should not be selected Incorrect. This hyperparameter has little impact and it is usually better to use the default values $0.9, 0.999$. | |
| 3. | Even if enough computational power is available for hyperparameter tuning, it is always better to babysit one model ("Panda" strategy), since this will result in a more custom model. True/False? | |
| | False | |
| | ○ True | |
| | Correct Correct. Although it is possible to create good models using the "Panda" strategy, obtaining better results is more likely using a "caviar" strategy due to the number of tests and the nature of the deep learning process of ideas, code, and experiment. | |
| 4. | If you think β (hyperparameter for momentum) is between 0.9 and 0.99, which of the following is the recommended way to sample a value for beta? | |
| | r = np.random.rand() beta = r*0.09 + 0.9 | |
| | | |
| | r = np.random.rand() beta = 1-10**(- r - 1) | |
| | r = np.random.rand() beta = r*0.9 + 0.09 | |
| | 0 | |
| | $r = np.random.rand()$ beta = $1-10^{**}(-r+1)$ | |
| | ⊘ Correct | |
| | | |
| 5. | Finding new values for the hyperparameters, once we have found good ones for a model, should only be done if new hardware or computational power is acquired. True/False? | |
| | False | |
| | ○ True | |
| | Correct Correct. As the data changes for the model, it might be beneficial to tune some of the hyperparameters again. | |

| 6. | When using batch normalization it is OK to drop the parameter $W^{[l]}$ from the forward propagation since it will be subtracted out when we compute $	ilde{z}^{[l]} = \gamma z_{	ext{normalize}}^{[l]} + eta^{[l]}$. True/False? |
|-----|---|
| | ○ True |
| | False |
| | $igodots$ Correct Correct. The parameter $W^{[l]}$ doesn't get subtracted during the batch normalization process, although it gets re-scaled. |
| 7. | Which of the following are true about batch normalization? |
| | One intuition behind why batch normalization works is that it helps reduce the internal covariance. |
| | $igcup$ The parameter ϵ in the batch normalization formula is used to accelerate the convergence of the model. |
| | $igcup$ The parameters eta and γ of batch normalization can't be trained using Adam or RMS prop. |
| | $igcup$ There is a global value of γ and eta that is used for all the hidden layers where batch normalization is used. |
| | Correct Yes. Internal covariance is a name to express that there has been a change in the distribution of the activations. Since after each iteration of gradient descent the parameters of a layer change, we might think that the activations suffer from covariance shift. |
| | gradient descent the parameters of a tayer change, we might dinik that the activations suffer from covariance sinit. |
| 8. | Which of the following is true about batch normalization? |
| | \bigcirc The optimal values to use for γ and eta are $\gamma=\sqrt{\sigma^2+\epsilon}$ and $eta=\mu.$ |
| | $igcirc z_{norm}^{(i)} = rac{z^{(i)}-\mu}{\sqrt{\sigma^2}}.$ |
| | $lacktriangle$ The parameters $\gamma^{[l]}$ and $eta^{[l]}$ set the variance and mean of $\widetilde{z}^{[l]}$. |
| | $igcap$ The parameters $\gamma^{[l]}$ and $eta^{[l]}$ can be learned only using plain gradient descent. |
| | $igotimes$ Correct. When applying the linear transformation $	ilde{z}^{(l)}=eta^{[l]}z^{(l)}_{norm}+\gamma^{[l]}$ we set the variance and mean of $	ilde{z}^{[l]}$. |
| 9. | A neural network is trained with Batch Norm. At test time, to evaluate the neural network on a new example you should perform the normalization using μ and σ^2 estimated using an exponentially weighted average across mini-batches seen during training. True/false? |
| | ○ False |
| | True |
| | \odot Correct Correct. This is a good practice to estimate the μ and σ^2 to use since at test time we might not be predicting over a batch of the same size, or it might even be a single example, thus using the μ and σ^2 of a single sample doesn't make sense. |
| 10. | Which of the following are some recommended criteria to choose a deep learning framework? |
| | It must run exclusively on cloud services, to ensure its robustness. |
| | Running speed. |
| | It must be implemented in C to be faster. |
| | It must use Python as the primary language. |
| | ○ Correct Correct. The running speed is a major factor, especially when working with large datasets. |
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