## Your grade: 90%

Next item ightarrow

Your latest: 90% • Your highest: 90% • To pass you need at least 80%. We keep your highest score.

1. With a relatively small set of hyperparameters, it is OK to use a grid search. True/False?

1/1 point

False

True

∠ Expand



Correct. When the set of hyperparameters is small like a range for  $n_l=1,2,3$  grid search works fine.

 $\epsilon$  in Adam.

0 / 1 point

You didn't select all the correct answers

During hyperparameter search, whether you try to babysit one model ("Panda" strategy) or train a lot of models in parallel ("Caviar") is largely determined by:

1/1 point

- The presence of local minima (and saddle points) in your neural network
- Whether you use batch or mini-batch optimization
- The number of hyperparameters you have to tune
- The amount of computational power you can access

Expand



✓ Correct

- 4. Knowing that the hyperparameter  $\alpha$  should be in the range of 0.00001 and 1.0, which of the following is the recommended way to sample a value for  $\alpha$ ?
  - r = -5\*np.random.rand()alpha = 10\*\*r
  - r = np.random.rand() alpha = 10\*\*r
  - r = np.random.rand() alpha = 0.00001 + r\*0.99999
  - r = -4\*np.random.rand() alpha = 10\*\*r

Expand

### ✓ Correct

Yes. This will generate a random value between  $10^{-5}$  and  $10^{0}$  chosen randomly in a logarithmic scale.

1/1 point

- 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  $\tilde{z}^{[l]} = \gamma z_{\mathrm{normalize}}^{[l]} + \beta^{[l]}$ . True/False?
  - False
  - ( ) True

Expand

✓ Correct

Correct. The parameter  $W^{[l]}$  doesn't get subtracted during the batch normalization process, although it gets re-scaled.

- 7. In the normalization formula  $z_{norm}^{(i)}=rac{z^{(i)}-\mu}{\sqrt{\sigma^2+arepsilon}}$  , why do we use epsilon? To avoid division by zero
  - To speed up convergence
    - In case  $\mu$  is too small
  - To have a more accurate normalization

# ∠ Expand



$$z_{norm}^{(i)} = rac{z^{(i)} - \mu}{\sqrt{\sigma^2}}$$
 .

- $\beta^{[l]}$  and  $\gamma^{[l]}$  are hyperparameters that must be tuned by random sampling in a logarithmic scale.
- The parameters  $\gamma^{[l]}$  and  $\beta^{[l]}$  set the variance and mean of  $\tilde{z}^{[l]}$ .

### ✓ Correct

Correct. When applying the linear transformation  $\tilde{z}^{(l)} = \beta^{[l]} z_{norm}^{(l)} + \gamma^{[l]}$  we set the variance and mean of  $\tilde{z}^{[l]}$ .

When using batch normalization we introduce two new parameters  $\gamma^{[l]}$ ,  $\beta^{[l]}$  that must be "learned" or trained.

### ✓ Correct

Correct. Batch normalization uses two parameters  $\beta$  and  $\gamma$  to compute  $\tilde{z}^{(i)}=\beta z_{norm}^{(i)}+\gamma$ .

Great, you got all the right answers.

After training a neural network with Batch Norm, at test time, to evaluate the neural network on a new example you should:

1/1 point

- Perform the needed normalizations, use  $\mu$  and  $\sigma^2$  estimated using an exponentially weighted average across mini-batches seen during training.
- Use the most recent mini-batch's value of  $\mu$  and  $\sigma^2$  to perform the needed normalizations.
- If you implemented Batch Norm on mini-batches of (say) 256 examples, then to evaluate on one test example, duplicate that example 256 times so that you're working with a minibatch the same size as during training.
- Skip the step where you normalize using  $\mu$  and  $\sigma^2$  since a single test example cannot be normalized.

Expand

1/1 point