### $\mathbf{A1}$

#### (a)

False. Deep neural networks have non-convex loss surfaces, so gradient descent does not guarantee the global optimum.

#### (b)

False. Initializing all weights to zero prevents breaking symmetry, causing identical updates and hindering training.

### (c)

True. Non-linear activation functions enable the network to learn non-linear decision boundaries, which would be impossible with purely linear transformations.

## (d)

False. Although the backward pass is more expensive than the forward pass, it is typically of the same order of magnitude and not prohibitively larger (big O time is the same).

# (e)

False. Neural networks are powerful and extensible, but they are not always the best choice for every circumstance due to factors like data requirements, computational cost, and interpretability.