

Figure 0.1: Example of a loss surface with two minima. If we place a hill at the position of the local minimum, it disappears and optimization will not get stuck.

# 1 Basics

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**Algorithm 1** Stochastic gradient descent

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**Require:** learning rate  $\lambda$

**Ensure:** a trained neural network

- 1: initialize the network, dataset and training parameters
  - 2: **while** stopping criteria is not met **do**
  - 3:   sample minibatch of  $m$  examples  $x^{(1)}, \dots, x^{(m)}$
  - 4:   compute gradient estimate  $\hat{g} = \frac{1}{m} \nabla_{\theta} \sum_i L(f(x^{(i)}; \theta), y^{(i)})$
  - 5:   apply parameter update  $\theta = \theta - \lambda \cdot \hat{g}$
  - 6: **end while**
  - 7: **return: the trained network**
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**Algorithm 2** Stochastic gradient descent with Momentum

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**Require:** learning rate  $\lambda$

**Require:** momentum parameter  $m$

**Ensure:** a trained neural network

- 1: initialize the network, dataset and training parameters
  - 2: **while** stopping criteria is not met **do**
  - 3:   sample minibatch of  $m$  examples  $x^{(1)}, \dots, x^{(m)}$
  - 4:   compute gradient estimate  $\hat{g} = \frac{1}{m} \nabla_{\theta} \sum_i L(f(x^{(i)}; \theta), y^{(i)})$
  - 5:   compute velocity update  $v = m \cdot v - \lambda \hat{g}$
  - 6:   apply parameter update  $\theta = \theta - v$
  - 7: **end while**
  - 8: **return: the trained network**
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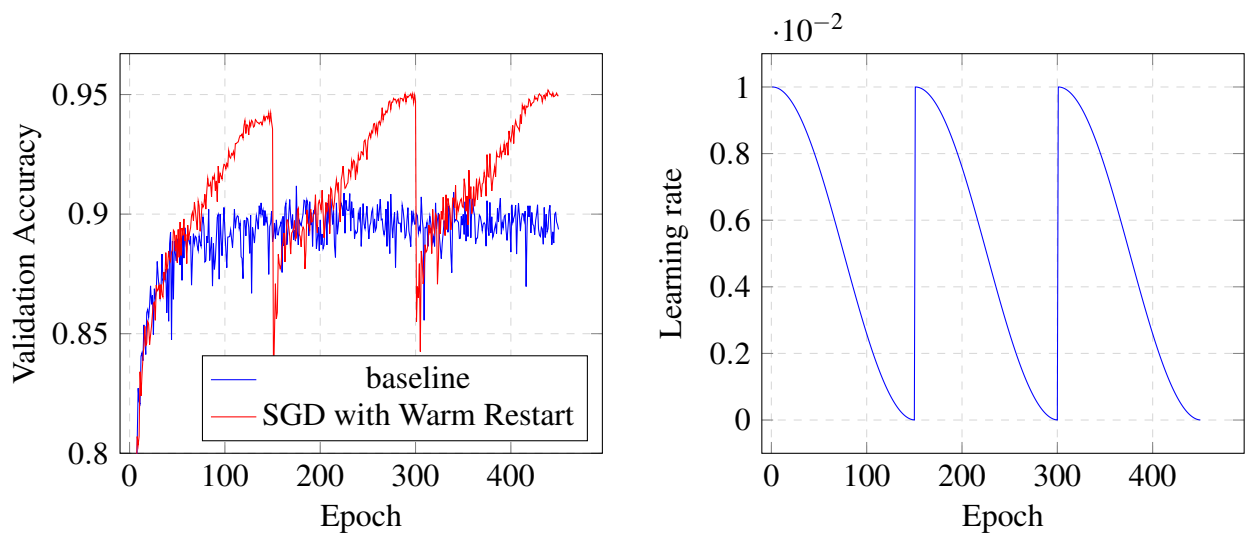


Figure 1.1: Cosine Decay with Warm Restart outperforms a fixed learning rate and increases the maximum accuracy for each restart (left). The right side shows how the learning rate is decayed.

## 2 Methods

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### Algorithm 3 Machine Learning with distancing

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**Require:** a set of parameters  $\theta$  and a dataset

**Ensure:** a assignment of  $\theta$  which maximizes performance

- 1: initialize the network, dataset and training parameters
  - 2: **for**  $i \leftarrow 1$  **to** desired number of epochs **do**
  - 3:   compute foward and backward pass of training data
  - 4:   update parameter values with optimizer
  - 5: **end for**
  - 6: create checkpoint we want to distance from
  - 7: **for**  $i \leftarrow$  next epoch **to** end **do**
  - 8:   **for** checkpoint **in** list of checkpoints **do**
  - 9:     compute parameter update which maintains performance but also increases distance to the checkpoint
  - 10:   **end for**
  - 11:   update parameter values with optimizer
  - 12: **end for**
  - 13: **return:** the final assignment of  $\theta$
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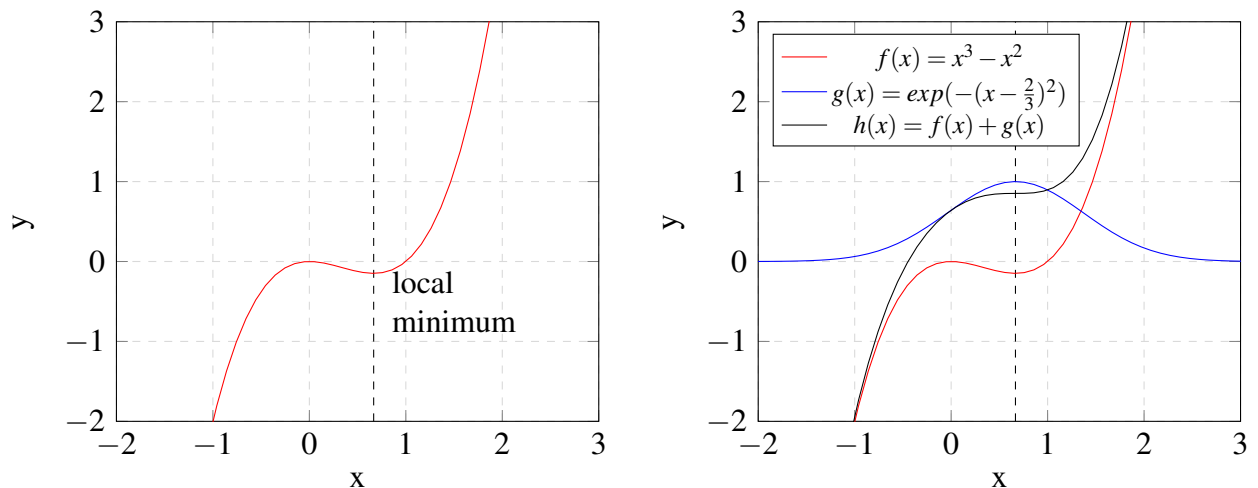


Figure 2.1: The red function has a local minimum at  $x = \frac{2}{3}$ . If we place a hill (blue) at this position and add the functions together the local minimum disappears (black).

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**Algorithm 4** Update step with distancing

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**Require:** learning rate  $\lambda$ , distance hyperparameters  $s$  and  $\sigma$

**Ensure:** a trained neural network

- 1: initialize the network, dataset and training parameters
  - 2: **while** stopping criteria is not met **do**
  - 3:   sample minibatch of  $m$  examples  $x^{(1)}, \dots, x^{(m)}$
  - 4:   compute gradient estimate  $\hat{g} = \nabla_{\theta} \frac{1}{m} \sum_i L(f(x^{(i)}; \theta), y^{(i)}) + s \cdot \frac{1}{c} \sum_c distance(\theta, \theta_c)$
  - 5:   apply parameter update  $\theta = \theta - \lambda \cdot \hat{g}$
  - 6: **end while**
  - 7: **return:** the trained network
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### 3 results

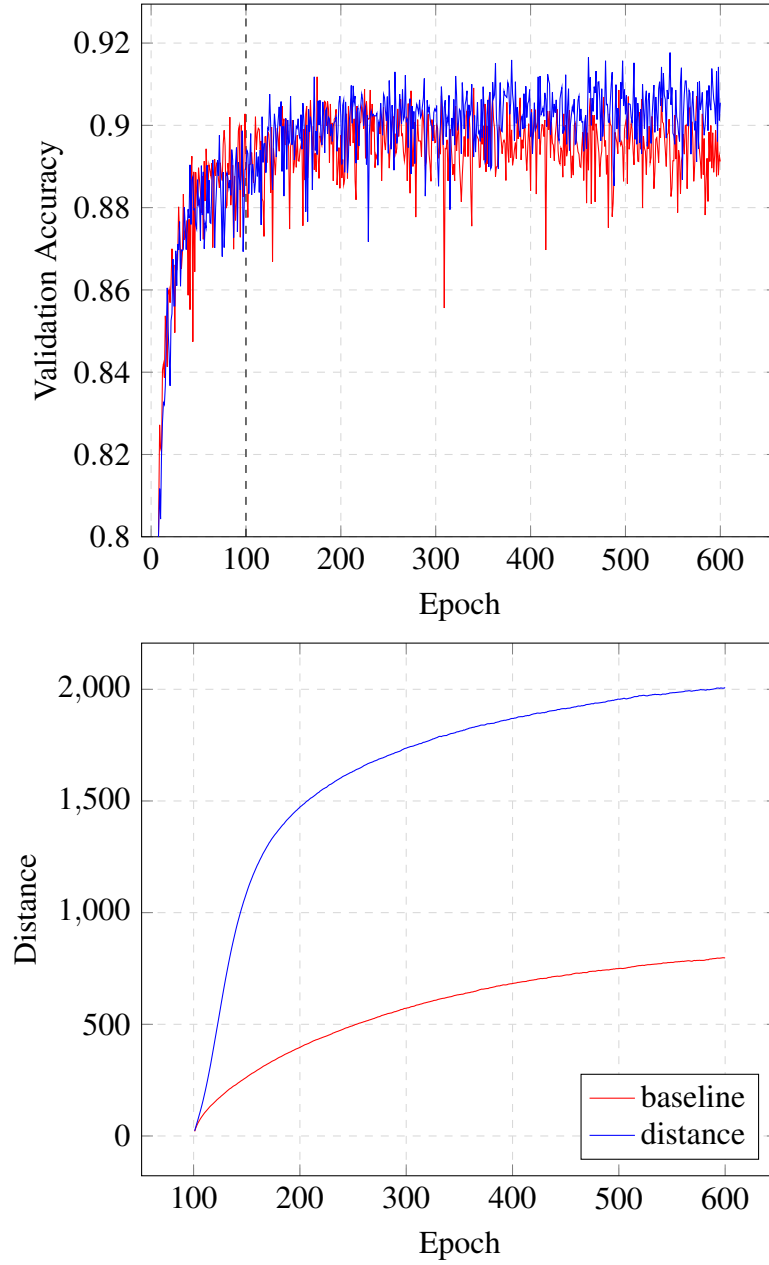


Figure 3.1: Baseline results for MobileNetV2. Upper plot shows validation accuracy, dotted line denotes addition of checkpoint. Lower plot shows  $L_2$  Distance to checkpoint.

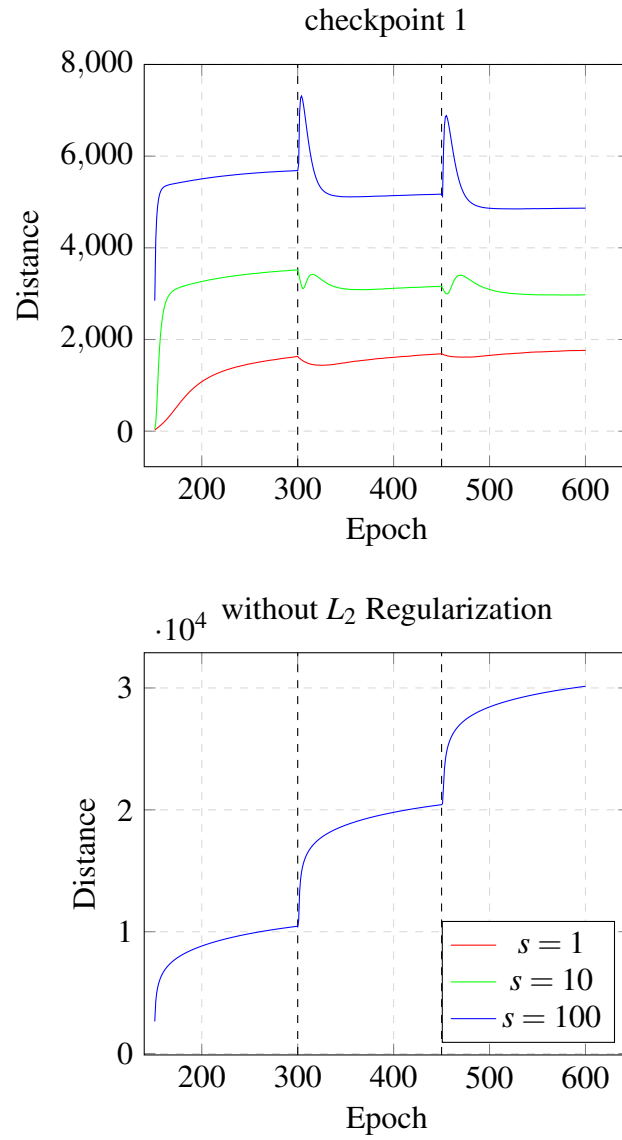


Figure 3.5: The upper plot shows the influence of new checkpoints on the distance to existing ones. Removing  $L_2$  Regularization changes this behaviour (lower plot).

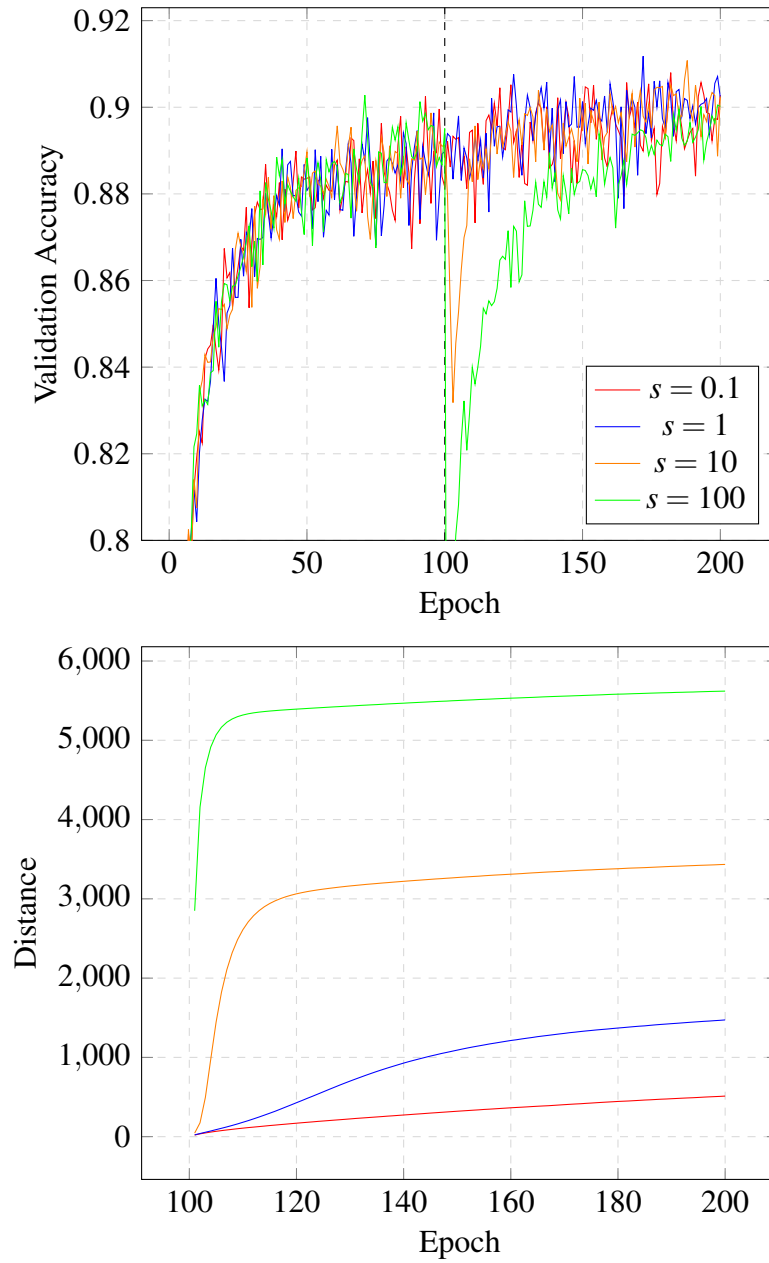


Figure 3.2: Influence of the strength hyperparameters  $s$  on the validation accuracy and distance.



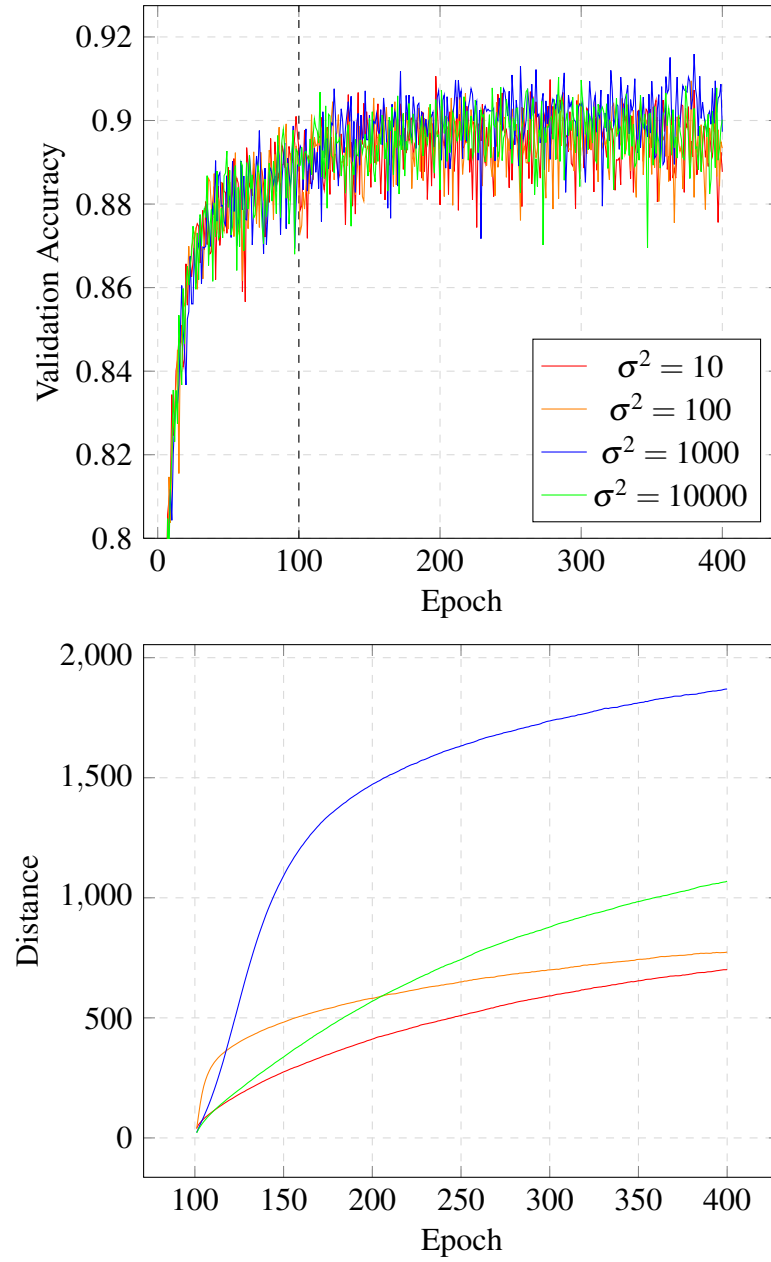


Figure 3.3: Influence of the width hyperparameters  $\sigma^2$  on the validation accuracy and distance.

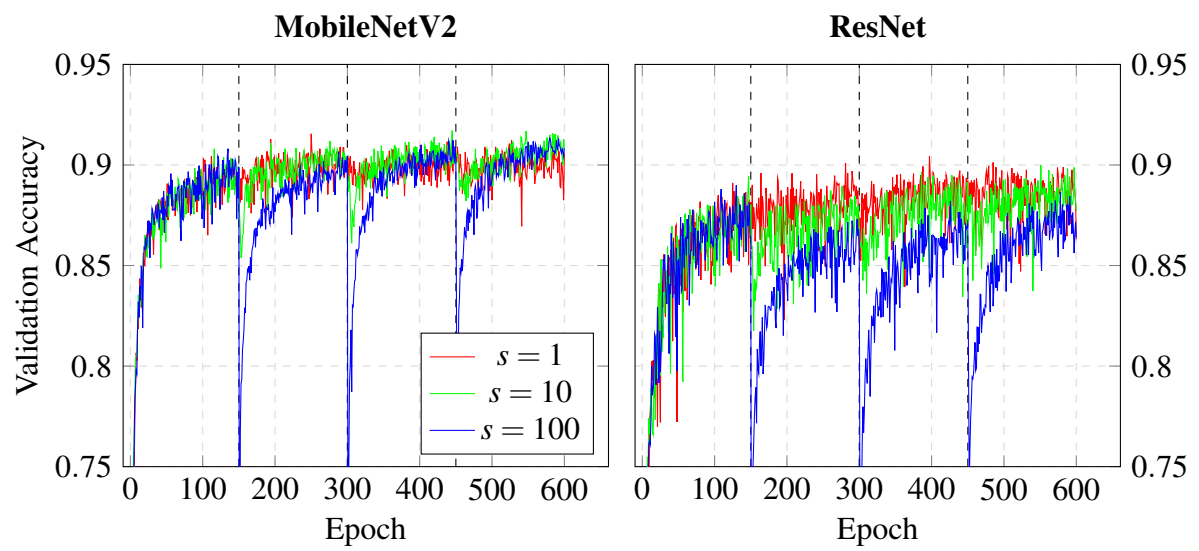


Figure 3.4: Influence of multiple checkpoints on the validation accuracy of MobileNetV2 and ResNet

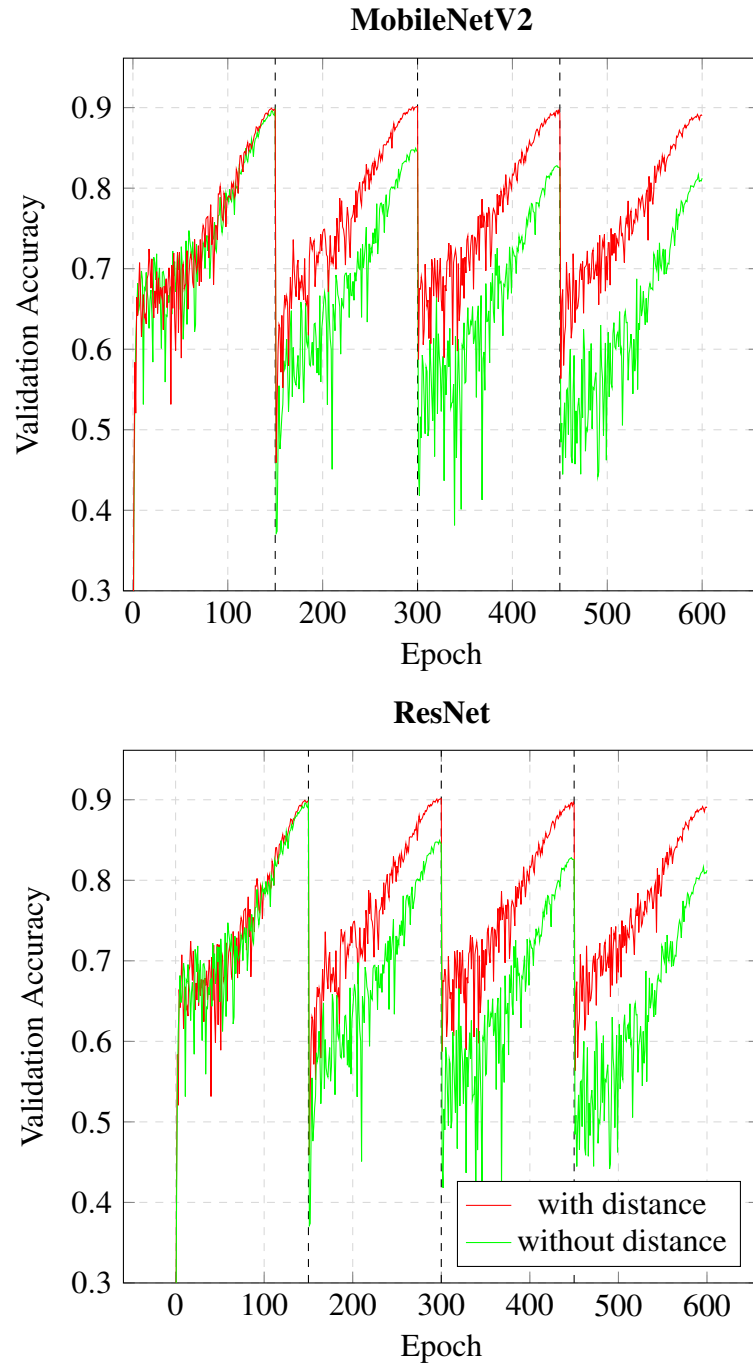


Figure 3.6: Influence of a suboptimal learning rate on Cosine Decay with Warm Restart without and with distance function.

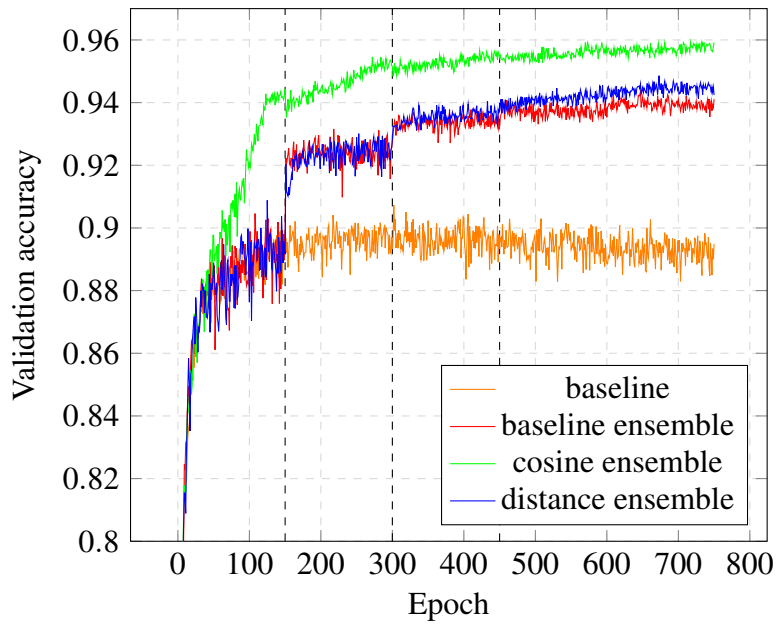


Figure 3.7: Ensemble accuracy for different networks against the baseline accuracy of a single network.

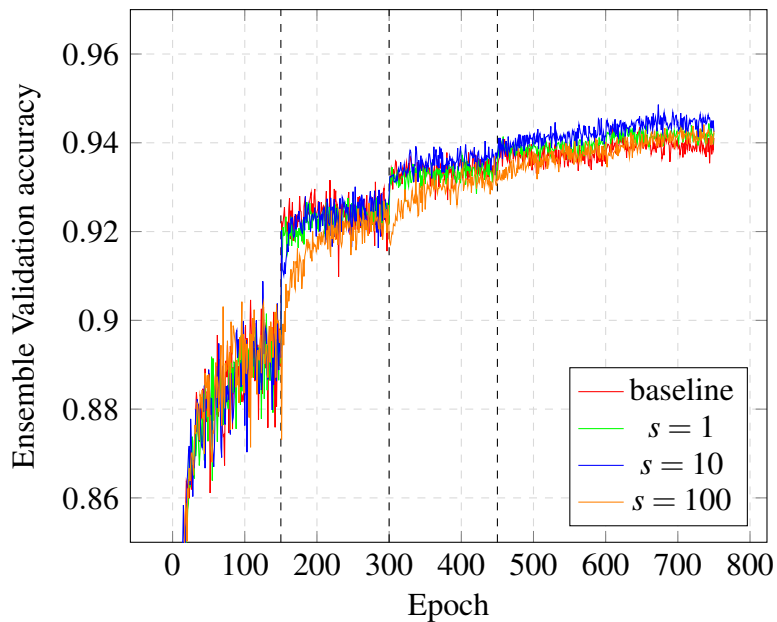


Figure 3.8: Ensemble accuracy for different strength values.

