

This is a modified output because the original had all 200 epochs.
I decided to show first and every 10th epoch.
I also added new line after train top_k_categorical_accuracy for aesthetics.
This is my cifar_100 case

Epoch 1/200
- 23s - loss: 4.5853 - acc: 0.0395 - top_k_categorical_accuracy: 0.1648
- val_loss: 4.2882 - val_acc: 0.0566 - val_top_k_categorical_accuracy: 0.2212

Epoch 10/200
- 19s - loss: 2.7005 - acc: 0.3017 - top_k_categorical_accuracy: 0.6256
- val_loss: 2.5983 - val_acc: 0.3295 - val_top_k_categorical_accuracy: 0.6479

Epoch 20/200
- 19s - loss: 2.2987 - acc: 0.3890 - top_k_categorical_accuracy: 0.7123
- val_loss: 2.1179 - val_acc: 0.4281 - val_top_k_categorical_accuracy: 0.7544

Epoch 30/200
- 19s - loss: 2.1717 - acc: 0.4196 - top_k_categorical_accuracy: 0.7422
- val_loss: 2.0597 - val_acc: 0.4579 - val_top_k_categorical_accuracy: 0.7620

Epoch 40/200
- 19s - loss: 1.9454 - acc: 0.4700 - top_k_categorical_accuracy: 0.7826 -
val_loss: 1.9816 - val_acc: 0.4687 - val_top_k_categorical_accuracy: 0.7816

Epoch 50/200
- 19s - loss: 1.8638 - acc: 0.4876 - top_k_categorical_accuracy: 0.7963
- val_loss: 1.8246 - val_acc: 0.5141 - val_top_k_categorical_accuracy: 0.8162

Epoch 60/200
- 19s - loss: 1.6748 - acc: 0.5325 - top_k_categorical_accuracy: 0.8314
- val_loss: 1.7033 - val_acc: 0.5431 - val_top_k_categorical_accuracy: 0.8296

Epoch 70/200
- 19s - loss: 1.7501 - acc: 0.5127 - top_k_categorical_accuracy: 0.8175
- val_loss: 1.7578 - val_acc: 0.5255 - val_top_k_categorical_accuracy: 0.8194

Epoch 80/200
- 19s - loss: 1.6369 - acc: 0.5421 - top_k_categorical_accuracy: 0.8380
- val_loss: 1.7103 - val_acc: 0.5449 - val_top_k_categorical_accuracy: 0.8276

Epoch 90/200
- 19s - loss: 1.6480 - acc: 0.5398 - top_k_categorical_accuracy: 0.8381
- val_loss: 1.7076 - val_acc: 0.5521 - val_top_k_categorical_accuracy: 0.8286

Epoch 100/200
- 19s - loss: 1.5067 - acc: 0.5706 - top_k_categorical_accuracy: 0.8604
- val_loss: 1.6648 - val_acc: 0.5603 - val_top_k_categorical_accuracy: 0.8390

Epoch 110/200
- 19s - loss: 1.5767 - acc: 0.5562 - top_k_categorical_accuracy: 0.8496
- val_loss: 1.7459 - val_acc: 0.5453 - val_top_k_categorical_accuracy: 0.8298

Epoch 120/200
- 19s - loss: 1.4045 - acc: 0.5976 - top_k_categorical_accuracy: 0.8795
- val_loss: 1.7573 - val_acc: 0.5541 - val_top_k_categorical_accuracy: 0.8338

Epoch 130/200
- 19s - loss: 1.3542 - acc: 0.6087 - top_k_categorical_accuracy: 0.8859
- val_loss: 1.7166 - val_acc: 0.5637 - val_top_k_categorical_accuracy: 0.8412

Epoch 140/200
- 19s - loss: 1.4385 - acc: 0.5940 - top_k_categorical_accuracy: 0.8706
- val_loss: 1.7157 - val_acc: 0.5663 - val_top_k_categorical_accuracy: 0.8306

Epoch 150/200
- 19s - loss: 1.4130 - acc: 0.6006 - top_k_categorical_accuracy: 0.8799
- val_loss: 1.8080 - val_acc: 0.5381 - val_top_k_categorical_accuracy: 0.8194

Epoch 160/200
- 19s - loss: 1.5302 - acc: 0.5681 - top_k_categorical_accuracy: 0.8564
- val_loss: 1.7696 - val_acc: 0.5629 - val_top_k_categorical_accuracy: 0.8374

Epoch 170/200
- 19s - loss: 1.4339 - acc: 0.5958 - top_k_categorical_accuracy: 0.8728
- val_loss: 1.7339 - val_acc: 0.5555 - val_top_k_categorical_accuracy: 0.8326

Epoch 180/200
- 19s - loss: 1.4378 - acc: 0.5932 - top_k_categorical_accuracy: 0.8708
- val_loss: 1.7565 - val_acc: 0.5663 - val_top_k_categorical_accuracy: 0.8372

Epoch 190/200
- 19s - loss: 1.4014 - acc: 0.6106 - top_k_categorical_accuracy: 0.8776

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- val_loss: 1.6688 - val_acc: 0.5647 - val_top_k_categorical_accuracy: 0.8304
Epoch 200/200
- 19s - loss: 1.2619 - acc: 0.6379 - top_k_categorical_accuracy: 0.8998
- val_loss: 1.7847 - val_acc: 0.5755 - val_top_k_categorical_accuracy: 0.8418
Test loss: 1.7742234039306641
Test accuracy: 0.5744
Test top 5 accuracy: 0.8374
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