## **EXPERIMENT 3**

Downloading data from <a href="https://www.cs.toronto.edu/~kriz/cifar-10-python.tar.gz">https://www.cs.toronto.edu/~kriz/cifar-10-python.tar.gz</a> 170498071/170498071 ————————————————————— 48 @us/step

Downloading data from <a href="https://www.cs.toronto.edu/~kri2/cifar-10-pytnon.tar.g2">https://www.cs.toronto.edu/~kri2/cifar-10-pytnon.tar.g2</a>
TP498071/170498071

Training data shape: (50000, 32, 32, 3)

Testing data shape: (10000, 32, 32, 3)

/tmp/ipython-input-5562023477.py:22: DeprecationWarning: Conversion of an array with ndim > 0 to a scalar is deprecated, and will error in future. Ensure you extraplititle(class\_names[int(y\_train[i])])











automobile









/usr/local/lib/python3.12/dist-packages/keras/src/layers/convolutional/base\_conv.py:113: UserWarning: Do not pass an `input\_shape`/`input\_dim` argument to a layer. super().\_\_init\_\_(activity\_regularizer=activity\_regularizer, \*\*kwargs)
Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 32, 32, 32)	896
max_pooling2d (MaxPooling2D)	(None, 16, 16, 32)	0
conv2d_1 (Conv2D)	(None, 16, 16, 64)	18,496
max pooling2d 1 (MaxPooling2D)	(None, 8, 8, 64)	9

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max_pooling2d_1 (MaxPooling2D)	(None, 8, 8, 64)	0
conv2d_2 (Conv2D)	(None, 8, 8, 128)	73,856
max_pooling2d_2 (MaxPooling2D)	(None, 4, 4, 128)	9
flatten (Flatten)	(None, 2048)	0
dense (Dense)	(None, 128)	262,272
dropout (Dropout)	(None, 128)	0
dense_1 (Dense)	(None, 10)	1,290

Total params: 356,810 (1.36 MB)
Trainable params: 356,810 (1.36 MB)

Non-trainable params: 0 (0	0.00 B)
Epoch 1/10	
782/782	— 111s 139ms/step - accuracy: 0.3200 - loss: 1.8397 - val_accuracy: 0.5391 - val_loss: 1.2878
Epoch 2/10	
782/782	— 139s 136ms/step - accuracy: 0.5309 - loss: 1.3165 - val_accuracy: 0.6295 - val_loss: 1.0834
Epoch 3/10	
782/782	— 109s 139ms/step - accuracy: 0.6062 - loss: 1.1187 - val_accuracy: 0.6742 - val_loss: 0.9176
Epoch 4/10	
782/782	— 107s 136ms/step - accuracy: 0.6603 - loss: 0.9751 - val_accuracy: 0.7066 - val_loss: 0.8459
Epoch 5/10	
782/782	— 106s 136ms/step - accuracy: 0.6949 - loss: 0.8814 - val_accuracy: 0.7172 - val_loss: 0.8090
Epoch 6/10	
782/782	- 108s 138ms/step - accuracy: 0.7216 - loss: 0.7975 - val_accuracy: 0.7360 - val_loss: 0.7663
Epoch 7/10	
782/782	- 141s 136ms/step - accuracy: 0.7445 - loss: 0.7442 - val_accuracy: 0.7453 - val_loss: 0.7497
Epoch 8/10	
782/782	- 107s 137ms/step - accuracy: 0.7629 - loss: 0.6849 - val_accuracy: 0.7554 - val_loss: 0.7205
Epoch 9/10	
782/782	— 106s 135ms/step - accuracy: 0.7818 - loss: 0.6281 - val_accuracy: 0.7557 - val_loss: 0.7289
Epoch 10/10	
782/782	— 106s 136ms/step - accuracy: 0.7922 - loss: 0.5946 - val_accuracy: 0.7438 - val_loss: 0.7779
Test Accuracy: 74.38%	
Test Loss: 0.7779	
1/1 0s	s 130ms/step

Epoch 4/10				
782/782	107s 136ms/step	- accuracy: 0.6603	- loss: 0.9751 - val_a	ccuracy: 0.7066 - val_loss: 0.8459
Epoch 5/10				
782/782	106s 136ms/step	- accuracy: 0.6949	- loss: 0.8814 - val_a	ccuracy: 0.7172 - val_loss: 0.8090
Epoch 6/10				
782/782	108s 138ms/step	- accuracy: 0.7216	- loss: 0.7975 - val_a	ccuracy: 0.7360 - val_loss: 0.7663
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Epoch 9/10				
782/782	106s 135ms/step	- accuracy: 0.7818	- loss: 0.6281 - val_a	ccuracy: 0.7557 - val_loss: 0.7289
Epoch 10/10				
	<b>106s</b> 136ms/step	- accuracy: 0.7922	- loss: 0.5946 - val_a	ccuracy: 0.7438 - val_loss: 0.7779
Test Accuracy: 74.38%				
Test Loss: 0.7779				
1/1	<b>– 0s</b> 130ms/step			
Pred: doa	Pred: ship	Pred: ship	Pred: airplane	Pred: frog
True: cat	True: ship	True: ship	True: airplane	True: frog









Pred: frog True: frog







Pred: automobile True: automobile

