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   "from tensorflow.keras.datasets import mnist\n",
   "from tensorflow.keras.models import Sequential\n",
   "from tensorflow.keras.layers import Conv2D, Dense, Flatten\n",
   "from tensorflow.keras.optimizers import Adam\n",
   "from tensorflow.keras.models import load model\n",
   "from PIL import Image, ImageOps\n",
   "import numpy"
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]

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  "Y test = np utils.to categorical(y test, number of classes)"
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  "model.add(Conv2D(32, (3, 3), activation=\"relu\"))\n",
  "model.add(Flatten())\n",
  "model.add(Dense(number of classes, activation=\"softmax\"))"
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       "Epoch 3/5\n",
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- val loss: 0.0906 - val accuracy: 0.9755\n",
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 "print(metrics)"
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 "prediction = model.predict(X_{test}[:4])\n",
 "print(prediction)"
],
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 "outputId": "13963cd5-c4d2-4dea-d645-19d11808c5c4"
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   "1/1 [=
                                                  =] - 0s 64ms/step\n",
   "[[4.77358049e-11\ 1.26020884e-14\ 2.23637656e-07\ 2.59297366e-07\n",
   " 1.53105145e-18 1.41474479e-13 2.73819453e-19 9.99999523e-01\n",
   " 5.75746352e-12 1.40723442e-08\\n",
```

```
"[3.92702641e-05 3.63764530e-09 9.99928832e-01 1.10518204e-06\n",
   " 3.28396650e-11 1.87219923e-13 3.02575540e-06 4.75269130e-12\n",
   " 2.79003762e-05 1.17118581e-09]\n",
   " [3.37602168e-11 9.99982953e-01 7.10459869e-09 3.63090309e-13\n",
   " 1.67968246e-05 6.36366426e-09 4.59948364e-11 2.65287614e-09\n",
   " 2.72516672e-07 1.53049936e-12]\n",
   "[9.99999762e-01 1.02759820e-17 6.89465485e-10 4.13503087e-14\n",
   " 3.53135576e-12 2.56500203e-11 6.89072754e-09 4.50628203e-14\n",
   " 8.74276596e-10 1.82247064e-07]]\n"
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 "print(numpy.argmax(prediction, axis=1))\n",
 "print(Y test[:4])"
],
"metadata": {
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   "[[0. 0. 0. 0. 0. 0. 0. 1. 0. 0.]\n",
   " [0. 0. 1. 0. 0. 0. 0. 0. 0. 0.]\n",
   " [0. 1. 0. 0. 0. 0. 0. 0. 0. 0.]\n",
   " [1. 0. 0. 0. 0. 0. 0. 0. 0. 0.]\n"
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 "model.save(\"model.h5\")"
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"execution count": 23,
"outputs": []
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"source": [
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```
"model=load_model(\"model.h5\")"
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}
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