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        tensorflow.keras.models import Sequential \n","from
        tensorflow.keras import layers \n",
        "from tensorflow.keras.layers import Dense, Flatten \n",
```

```
"from tensorflow.keras.layers import Conv2D \n", "from keras.
  utils import np utils \n",
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   (60000, 28, 28, 1).astype('float32')\n",
  "x test=x test.reshape (10000, 28, 28, 1).astype ('float32')\n","\n",
  "x train.shape"
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          "model.add(Conv2D(64, (3, 3), input shape=(28, 28, 1),
activation='relu'))\n",
           "model.add(Conv2D(32, (3, 3), activation = 'relu'))\n","\n",
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optimizer=\"Adam\", metrics=['accuracy'])"
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- loss: 0.2126 - accuracy: 0.9534 - val loss: 0.0857 - val accuracy: 0.9727\n",
                "Epoch 2/5\n",
                "1875/1875 [=
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- loss: 0.0633 - accuracy: 0.9806 - val loss: 0.0812 - val accuracy: 0.9751\n",
                "Epoch 3/5\n",
                "1875/1875 [=
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- loss: 0.0451 - accuracy: 0.9863 - val loss: 0.0776 - val accuracy: 0.9780\n",
                "Epoch 4/5\n",
                "1875/1875 [=
                                                                         =] - 122s 65ms/step
- loss: 0.0313 - accuracy: 0.9905 - val loss: 0.1089 - val accuracy: 0.9736\n",
                "Epoch 5/5\n".
```

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- loss: 0.0277 - accuracy: 0.9917 - val loss: 0.0893 - val accuracy:0.9779\n"
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9.9999857e-01]]\n"
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          "import numpy as np\n", "print(np.argmax(prediction,
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