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     "print(y_test. shape)\n",
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     "print(scores) # returns loss and accuracy"
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"\n".
"predictions = model. predict(x_test)\n".
"# Convert prediction to labels\n".
"label=[]\n".
"for pred in predictions:\n".
"max_val = max(pred)\n".
"for (i, p) in enumerate(pred):"
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