Requirement already satisfied: p 0) Requirement already satisfied: m w) (0.1.2) [notice] A new release of pip is [notice] To update, run: python.	ygments<3.0.0,>=2.13.0 in c:\users\j durl~=0.1 in c:\users\jadal\appdata\ available: 24.0 -> 24.1.1	jadal\appdata\roaming\	<pre>hpython\python312\site-pa</pre>	ckages (from rich->keras>=3.0.0	tensorflow-intel==2.16.2->tensorflow) (3>tensorflow-intel==2.16.2->tensorflow) (ras>=3.0.0->tensorflow-intel==2.16.2->ten
<pre>import pandas as pd import numpy as np import seaborn as sns from tensorflow import keras df = pd.read_csv('Irisdataset.com')</pre>	sv')				
df.head() Id SepalLengthCm SepalWidthCm 0 1 5.1 3.5 1 2 4.9 3.0	1.4 0.2 Iris-setosa 1.4 0.2 Iris-setosa	-			
2 3 4.7 3.2 3 4 4.6 3.1 4 5 5.0 3.6	1.5 0.2 Iris-setosa				
df['Species'].value_counts() Species Iris-setosa 50 Iris-versicolor 50 Iris-virginica 50 Name: count, dtype: int64					
df.info() <class #="" 'pandas.core.frame.datafr="" (total="" 0="" 14="" 150="" 6="" column="" columns="" columns):="" cou="" data="" entries,="" id="" non-null="" non-null<="" rangeindex:="" td="" to=""><td>nt Dtype</td><td></td><td></td><td></td><td></td></class>	nt Dtype				
1 SepalLengthCm 150 non-null 2 SepalWidthCm 150 non-null 3 PetalLengthCm 150 non-null 4 PetalWidthCm 150 non-null 5 Species 150 non-null dtypes: float64(4), int64(1), ob memory usage: 7.2+ KB	float64 float64 float64 float64 object				
df.isnull().sum() Id 0 SepalLengthCm 0 SepalWidthCm 0 PetalLengthCm 0 PetalWidthCm 0					
Species 0 dtype: int64 from sklearn.preprocessing import le = LabelEncoder() df['Species'] = le.fit_transfor df.head()					
0 1 5.1 3.5 1 2 4.9 3.0	1.4 0.2 0				
2 3 4.7 3.2 3 4 4.6 3.1 4 5 5.0 3.6 Species_name = le.classes_	1.5 0.2 0				
<pre>print(species_name) ['Iris-setosa' 'Iris-versicolor' X = df.drop(columns=['Id', 'Spey = df['Species'] X.head(3)</pre>					
SepalLengthCm SepalWidthCm Per 0 5.1 3.5 1 4.9 3.0 2 4.7 3.2	1.4 0.2 1.4 0.2 1.3 0.2				
<pre>print(y[:5]) 0 0 1 0 2 0</pre>					
3 0 4 0 Name: Species, dtype: int32 from sklearn.model_selection im x_train, x_test, y_train, y_tes x_train.shape	<pre>port train_test_split t = train_test_split(X, y, test_size</pre>	e=0.3, shuffle =True ,	random_state=69)		
<pre>from sklearn.preprocessing impos sc = StandardScaler() print(x_train[:1]) x_train = sc.fit_transform(x_tr</pre>					
<pre>x_test = sc.transform(x_test) print(x_train[:1]) SepalLengthCm SepalWidthCm 57</pre>	PetalLengthCm PetalWidthCm 3.3 1.0 50655 -0.258263]]				
print(y_train[:5]) [[0. 1. 0.] [0. 0. 1.] [0. 1. 0.] [0. 1. 0.] [0. 1. 0.]	Tour(y_truin, num_orusses=s)				
from keras.models import Sequer from keras.layers import Dense, from tensorflow.keras.models im from tensorflow.keras.layers im # Define the model	Dropout port Sequential				
<pre>model = Sequential() # Add the input layer model.add(Input(shape=(x_train. # Add subsequent layers model.add(Dense(units=32, active))</pre>	ation='relu'))				
<pre>model.add(Dense(units=32, active model.add(Dropout(0.5)) model.add(Dense(units=3, active))</pre>	ation='relu'))	etrics=['accuracy'])			
dense_3 (Dense) dense_4 (Dense)	(None, 32)		056	 	
dropout_1 (Dropout) dense_5 (Dense) Total params: 1,315 (5.14 KB) Trainable params: 1,315 (5.14 Non-trainable params: 0 (0.00	•		99		
Epoch 1/100 4/4 - 2s - 623ms/step - accuracy Epoch 2/100 4/4 - 0s - 15ms/step - accuracy:	: 0.3619 - loss: 1.0886				
Epoch 3/100 4/4 - 0s - 16ms/step - accuracy: Epoch 4/100 4/4 - 0s - 17ms/step - accuracy: Epoch 5/100 4/4 - 0s - 41ms/step - accuracy: Epoch 6/100 4/4 - 0s - 14ms/step - accuracy:	0.5048 - loss: 0.9349 0.6000 - loss: 0.8759				
Epoch 7/100 4/4 - 0s - 14ms/step - accuracy: Epoch 8/100 4/4 - 0s - 12ms/step - accuracy: Epoch 9/100 4/4 - 0s - 12ms/step - accuracy: Epoch 10/100	0.6286 - loss: 0.7829 0.6286 - loss: 0.7359				
4/4 - 0s - 11ms/step - accuracy: Epoch 11/100 4/4 - 0s - 11ms/step - accuracy: Epoch 12/100 4/4 - 0s - 11ms/step - accuracy: Epoch 13/100 4/4 - 0s - 12ms/step - accuracy: Epoch 14/100	0.6571 - loss: 0.6996 0.6952 - loss: 0.6523				
4/4 - 0s - 12ms/step - accuracy: Epoch 15/100 4/4 - 0s - 12ms/step - accuracy: Epoch 16/100 4/4 - 0s - 21ms/step - accuracy: Epoch 17/100 4/4 - 0s - 9ms/step - accuracy:	0.7619 - loss: 0.5928 0.7429 - loss: 0.5621				
Epoch 18/100 4/4 - 0s - 19ms/step - accuracy: Epoch 19/100 4/4 - 0s - 14ms/step - accuracy: Epoch 20/100 4/4 - 0s - 13ms/step - accuracy: Epoch 21/100 4/4 - 0s - 19ms/step - accuracy:	0.7810 - loss: 0.5397 0.8095 - loss: 0.5224				
Epoch 22/100 4/4 - 0s - 14ms/step - accuracy: Epoch 23/100 4/4 - 0s - 19ms/step - accuracy: Epoch 24/100 4/4 - 0s - 20ms/step - accuracy: Epoch 25/100	0.8000 - loss: 0.5025 0.7429 - loss: 0.5117				
4/4 - 0s - 19ms/step - accuracy: Epoch 26/100 4/4 - 0s - 13ms/step - accuracy: Epoch 27/100 4/4 - 0s - 13ms/step - accuracy: Epoch 28/100 4/4 - 0s - 39ms/step - accuracy:	0.8667 - loss: 0.4422 0.8190 - loss: 0.4587				
Epoch 29/100 4/4 - 0s - 18ms/step - accuracy: Epoch 30/100 4/4 - 0s - 11ms/step - accuracy: Epoch 31/100 4/4 - 0s - 12ms/step - accuracy: Epoch 32/100 4/4 - 0s - 25ms/step - accuracy:	0.8000 - loss: 0.4220 0.8857 - loss: 0.3801				
Epoch 33/100 4/4 - 0s - 15ms/step - accuracy: Epoch 34/100 4/4 - 0s - 21ms/step - accuracy: Epoch 35/100 4/4 - 0s - 11ms/step - accuracy: Epoch 36/100 4/4 - 0s - 16ms/step - accuracy:	0.8762 - loss: 0.3580 0.8476 - loss: 0.4249				
Epoch 37/100 4/4 - 0s - 17ms/step - accuracy: Epoch 38/100 4/4 - 0s - 13ms/step - accuracy: Epoch 39/100 4/4 - 0s - 11ms/step - accuracy: Epoch 40/100	0.8857 - loss: 0.3375 0.8095 - loss: 0.3776 0.8381 - loss: 0.3386				
4/4 - 0s - 11ms/step - accuracy: Epoch 41/100 4/4 - 0s - 18ms/step - accuracy: Epoch 42/100 4/4 - 0s - 13ms/step - accuracy: Epoch 43/100 4/4 - 0s - 12ms/step - accuracy:	0.8571 - loss: 0.3451 0.8571 - loss: 0.3304				
Epoch 44/100 4/4 - 0s - 20ms/step - accuracy: Epoch 45/100 4/4 - 0s - 15ms/step - accuracy: Epoch 46/100 4/4 - 0s - 12ms/step - accuracy: Epoch 47/100 4/4 - 0s - 12ms/step - accuracy:	0.8762 - loss: 0.3050 0.9143 - loss: 0.2928				
Epoch 48/100 4/4 - 0s - 23ms/step - accuracy: Epoch 49/100 4/4 - 0s - 13ms/step - accuracy: Epoch 50/100 4/4 - 0s - 14ms/step - accuracy: Epoch 51/100	0.9048 - loss: 0.3005 0.9143 - loss: 0.2638				
4/4 - 0s - 12ms/step - accuracy: Epoch 52/100 4/4 - 0s - 10ms/step - accuracy: Epoch 53/100 4/4 - 0s - 11ms/step - accuracy: Epoch 54/100 4/4 - 0s - 16ms/step - accuracy:	0.9333 - loss: 0.2690 0.9143 - loss: 0.2722				
Epoch 55/100 4/4 - 0s - 14ms/step - accuracy: Epoch 56/100 4/4 - 0s - 23ms/step - accuracy: Epoch 57/100 4/4 - 0s - 12ms/step - accuracy: Epoch 58/100 4/4 - 0s - 12ms/step - accuracy:	0.9048 - loss: 0.2436 0.9238 - loss: 0.2409				
Epoch 59/100 4/4 - 0s - 13ms/step - accuracy: Epoch 60/100 4/4 - 0s - 11ms/step - accuracy: Epoch 61/100 4/4 - 0s - 12ms/step - accuracy: Epoch 62/100	0.9238 - loss: 0.2349 0.9238 - loss: 0.2632				
4/4 - 0s - 12ms/step - accuracy: Epoch 63/100 4/4 - 0s - 11ms/step - accuracy: Epoch 64/100 4/4 - 0s - 13ms/step - accuracy: Epoch 65/100 4/4 - 0s - 12ms/step - accuracy: Epoch 66/100	0.9048 - loss: 0.2661 0.9333 - loss: 0.2308				
4/4 - 0s - 12ms/step - accuracy: Epoch 67/100 4/4 - 0s - 11ms/step - accuracy: Epoch 68/100 4/4 - 0s - 12ms/step - accuracy: Epoch 69/100 4/4 - 0s - 12ms/step - accuracy:	0.9238 - loss: 0.2198 0.9524 - loss: 0.1963				
Epoch 70/100 4/4 - 0s - 11ms/step - accuracy: Epoch 71/100 4/4 - 0s - 11ms/step - accuracy: Epoch 72/100 4/4 - 0s - 11ms/step - accuracy: Epoch 73/100 4/4 - 0s - 11ms/step - accuracy:	0.9333 - loss: 0.2145 0.9143 - loss: 0.2162				
Epoch 74/100 4/4 - 0s - 12ms/step - accuracy: Epoch 75/100 4/4 - 0s - 12ms/step - accuracy: Epoch 76/100 4/4 - 0s - 12ms/step - accuracy: Epoch 77/100	0.9143 - loss: 0.2375 0.9619 - loss: 0.1992 0.9619 - loss: 0.1661				
4/4 - 0s - 12ms/step - accuracy: Epoch 78/100 4/4 - 0s - 20ms/step - accuracy: Epoch 79/100 4/4 - 0s - 13ms/step - accuracy: Epoch 80/100 4/4 - 0s - 12ms/step - accuracy: Epoch 81/100	0.9333 - loss: 0.1878 0.9238 - loss: 0.1761				
4/4 - 0s - 10ms/step - accuracy: Epoch 82/100 4/4 - 0s - 12ms/step - accuracy: Epoch 83/100 4/4 - 0s - 8ms/step - accuracy: Epoch 84/100 4/4 - 0s - 12ms/step - accuracy:	0.9429 - loss: 0.1585 0.9524 - loss: 0.1703				
Epoch 85/100 4/4 - 0s - 21ms/step - accuracy: Epoch 86/100 4/4 - 0s - 16ms/step - accuracy: Epoch 87/100 4/4 - 0s - 15ms/step - accuracy: Epoch 88/100 4/4 - 0s - 16ms/step - accuracy:	0.9333 - loss: 0.1749 0.9524 - loss: 0.1952 0.9333 - loss: 0.1849				
Epoch 89/100 4/4 - 0s - 12ms/step - accuracy: Epoch 90/100 4/4 - 0s - 13ms/step - accuracy: Epoch 91/100 4/4 - 0s - 15ms/step - accuracy: Epoch 92/100	0.9524 - loss: 0.1513 0.9524 - loss: 0.1632 0.9619 - loss: 0.1529				
4/4 - 0s - 14ms/step - accuracy: Epoch 93/100 4/4 - 0s - 13ms/step - accuracy: Epoch 94/100 4/4 - 0s - 11ms/step - accuracy: Epoch 95/100 4/4 - 0s - 17ms/step - accuracy:	0.9619 - loss: 0.1793 0.9429 - loss: 0.1789				
Epoch 96/100 4/4 - 0s - 12ms/step - accuracy: Epoch 97/100 4/4 - 0s - 17ms/step - accuracy: Epoch 98/100 4/4 - 0s - 12ms/step - accuracy: Epoch 99/100 4/4 - 0s - 10ms/step - accuracy:	0.9333 - loss: 0.1740 0.9619 - loss: 0.1420				
<pre>4/4 - 0s - 10ms/step - accuracy: Epoch 100/100 4/4 - 0s - 9ms/step - accuracy:]: <keras.src.callbacks.history.hi]:="" prediction="model.predict(x_teprint(prediction[:5])</pre"></keras.src.callbacks.history.hi></pre>	0.9714 - loss: 0.1456 story at 0x1fb85ba8e30>				
2/2 [[9.99966621e-01 1.34209477e-05 [5.64658176e-03 9.26387787e-01 [9.99868751e-01 7.82671996e-05 [1.00378515e-04 9.11357068e-03	6.79656714e-02] 5.29615136e-05] 9.90786076e-01] 8.59556167e-05]]				
[9.99792755e-01 1.21366196e-04	on, axis=-1)				
<pre>prediction = np.argmax(predicti print(prediction[:5]) [0 1 0 2 0]</pre>					
<pre>prediction = np.argmax(predicti print(prediction[:5]) [0 1 0 2 0] print(y_test[:5]) 4 0 58 1 20 0 114 2 31 0 Name: Species, dtype: int32 from sklearn.metrics import according.</pre>					
prediction = np.argmax(prediction print(prediction[:5]) [0 1 0 2 0] [0 1 0 2 0] print(y_test[:5]) 4	ediction))				
prediction = np.argmax(prediction print(prediction[:5]) [0 1 0 2 0] [0 1 0 2 0] print(y_test[:5]) 4	ediction)) rediction) , fmt='d', cmap='Reds', xticklabels: for irsis dataset prediction') tsize=14)	=species_name, ytickl	abels=species_name)		

Iris-versicolor

prediction

lris-virginica

lris-setosa

In []:

In [1]: !pip install tensorflow

ow-intel==2.16.2->tensorflow) (4.25.3)

Defaulting to user installation because normal site-packages is not writeable

Requirement already satisfied: tensorflow in c:\users\jadal\appdata\roaming\python\python312\site-packages (2.16.2)

Requirement already satisfied: tensorflow-intel==2.16.2 in c:\users\jadal\appdata\roaming\python\python312\site-packages (from tensorflow) (2.16.2)

Requirement already satisfied: absl-py>=1.0.0 in c:\users\jadal\appdata\roaming\python\python312\site-packages (from tensorflow-intel==2.16.2->tensorflow) (2.1.0) Requirement already satisfied: astunparse>=1.6.0 in c:\users\jadal\appdata\roaming\python\python312\site-packages (from tensorflow-intel==2.16.2->tensorflow) (1.6.3) Requirement already satisfied: flatbuffers>=23.5.26 in c:\users\jadal\appdata\roaming\python\python312\site-packages (from tensorflow-intel==2.16.2->tensorflow) (24.3.25)

Requirement already satisfied: h5py>=3.10.0 in c:\users\jadal\appdata\roaming\python\python312\site-packages (from tensorflow-intel==2.16.2->tensorflow) (3.11.0) Requirement already satisfied: libclang>=13.0.0 in c:\users\jadal\appdata\roaming\python\python312\site-packages (from tensorflow-intel==2.16.2->tensorflow) (18.1.1) Requirement already satisfied: ml-dtypes~=0.3.1 in c:\users\jadal\appdata\roaming\python\python312\site-packages (from tensorflow-intel==2.16.2->tensorflow) (0.3.2) Requirement already satisfied: opt-einsum>=2.3.2 in c:\users\jadal\appdata\roaming\python\python312\site-packages (from tensorflow-intel==2.16.2->tensorflow) (3.3.0)

Requirement already satisfied: packaging in c:\users\jadal\appdata\roaming\python\python312\site-packages (from tensorflow-intel==2.16.2->tensorflow) (24.1)

Requirement already satisfied: setuptools in c:\users\jadal\appdata\roaming\python\python312\site-packages (from tensorflow-intel==2.16.2->tensorflow) (70.1.0) Requirement already satisfied: six>=1.12.0 in c:\users\jadal\appdata\roaming\python\python312\site-packages (from tensorflow-intel==2.16.2->tensorflow) (1.16.0) Requirement already satisfied: termcolor>=1.1.0 in c:\users\jadal\appdata\roaming\python\python312\site-packages (from tensorflow-intel==2.16.2->tensorflow) (2.4.0)

Requirement already satisfied: wrapt>=1.11.0 in c:\users\jadal\appdata\roaming\python\python312\site-packages (from tensorflow-intel==2.16.2->tensorflow) (1.16.0)

Requirement already satisfied: google-pasta>=0.1.1 in c:\users\jadal\appdata\roaming\python\python312\site-packages (from tensorflow-intel==2.16.2->tensorflow) (0.2.0)

Requirement already satisfied: requests<3,>=2.21.0 in c:\users\jadal\appdata\roaming\python\python312\site-packages (from tensorflow-intel==2.16.2->tensorflow) (2.32.3)

Requirement already satisfied: grpcio<2.0,>=1.24.3 in c:\users\jadal\appdata\roaming\python\python312\site-packages (from tensorflow-intel==2.16.2->tensorflow) (1.64.1) Requirement already satisfied: tensorboard<2.17,>=2.16 in c:\users\jadal\appdata\roaming\python\python312\site-packages (from tensorflow-intel==2.16.2->tensorflow) (2.16.2)

Requirement already satisfied: typing-extensions>=3.6.6 in c:\users\jadal\appdata\roaming\python\python312\site-packages (from tensorflow-intel==2.16.2->tensorflow) (4.12.2)

Requirement already satisfied: gast!=0.5.0,!=0.5.1,!=0.5.2,>=0.2.1 in c:\users\jadal\appdata\roaming\python\python312\site-packages (from tensorflow-intel==2.16.2->tensorflow) (0.6.0)

Requirement already satisfied: protobuf!=4.21.0,!=4.21.1,!=4.21.2,!=4.21.3,!=4.21.4,!=4.21.5,<5.0.0dev,>=3.20.3 in c:\users\jadal\appdata\roaming\python\python312\site-packages (from tensorflement already satisfied: protobuf!=4.21.0,!=4.21.1,!=4.21.2,!=4.21.3,!=4.21.4,!=4.21.5,<5.0.0dev,>=3.20.3 in c:\users\jadal\appdata\roaming\python\python\python312\site-packages (from tensorflement already satisfied: protobuf!=4.21.0,!=4.21.1,!=4.21.2,!=4.21.2,!=4.21.3,!=