```
from flask import Flask, render template, request
from tensorflow.keras.models import load model
from tensorflow.keras.preprocessing import image
import numpy as np
import os
app = Flask( name )
model = load model("model/poultry model.h5")
# Match the class labels used during training
class labels = ['Coccidiosis', 'New Castle', 'Healthy', 'salmonella'] #
Replace with your real class folder names
@app.route('/')
def index():
    return render template('index.html')
@app.route('/predict', methods=['POST'])
def predict():
    if 'file' not in request.files:
        return "No file part"
    file = request.files['file']
    if file.filename == '':
        return "No selected file"
    if file:
        file path = os.path.join('static/uploaded', file.filename)
        file.save(file path)
        # Preprocess the image
        img = image.load img(file path, target size=(224, 224))
        img array = image.img to array(img) / 255.0
        img array = np.expand dims(img array, axis=0)
        # Predict
        prediction = model.predict(img array)
        predicted class = class labels[np.argmax(prediction)]
        return render template('result.html', prediction=predicted class,
image path=file path)
if __name_ == ' main ':
    app.run (debug=True)
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   "import os\n",
```

```
"import numpy as np\n",
    "import tensorflow as tf\n",
    "from tensorflow.keras.preprocessing.image import
ImageDataGenerator\n",
    "from tensorflow.keras.applications import MobileNetV2\n",
    "from tensorflow.keras.models import Model, load model\n",
    "from tensorflow.keras.layers import Dense, GlobalAveragePooling2D\n",
    "from tensorflow.keras.optimizers import Adam\n",
    "from sklearn.metrics import classification report,
confusion matrix\n",
    "import cv2\n"
   ]
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    "base dir = 'dataset' # Keep your dataset here\n",
    "img height, img width = 224, 224\n",
    "batch size = 32\n",
    "epochs = 10\n"
   1
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     "Found 5451 images belonging to 4 classes.\n",
      "Found 1361 images belonging to 4 classes.\n"
     1
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   ],
   "source": [
    "train datagen = ImageDataGenerator(\n",
         rescale=1./255,\n",
                                        # 80% train, 20% validation\n",
         validation split=0.2,
    "
         horizontal flip=True, \n",
         zoom range=0.2,\n",
         shear range=0.2\n",
    ")\n",
    "\n",
    "train generator = train_datagen.flow_from_directory(\n",
        base dir, \n",
        target_size=(img_height, img width), \n",
        batch size=batch size, \n",
```

```
subset='training', \n",
         class mode='categorical'\n",
    ")\n",
    "\n",
    "val generator = train datagen.flow from directory(\n",
         base dir, \n",
         target size=(img height, img width), \n",
    11
         batch size=batch size, \n",
         subset='validation', \n",
         class mode='categorical'\n",
    ")\n"
   ]
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    "base model = MobileNetV2(input shape=(img height, img width, 3),
include top=False, weights='imagenet') \n",
    "base model.trainable = False # Freeze pre-trained layers\n",
    "\n",
    "x = base model.output\n",
    "x = GlobalAveragePooling2D()(x)\n",
    "x = Dense(128, activation='relu')(x)\n",
    "predictions = Dense(train generator.num classes,
activation='softmax')(x)n",
    "\n",
    "model = Model(inputs=base model.input, outputs=predictions) n"
   1
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   "source": [
    "model.compile(optimizer=Adam(learning rate=0.0001), \n",
                   loss='categorical crossentropy', \n",
                   metrics=['accuracy']) \n"
   ]
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```

```
"text": [
      "C:\\Users\\91939\\anaconda3\\Lib\\site-
packages\\keras\\src\\trainers\\data adapters\\py dataset adapter.py:121:
UserWarning: Your `PyDataset` class should call
`super().__init__(**kwargs)` in its constructor. `**kwargs` can include
`workers`, `use multiprocessing`, `max queue size`. Do not pass these
arguments to `fit()`, as they will be ignored.\n",
      " self. warn if super not called()\n"
    },
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      "Epoch 1/10\n",
      "\u001b[1m171/171\u001b[0m \u001b[32m-
 -\u001b[0m\u001b[37m\u001b[0m \u001b[1m1743s\u001b[0m 10s/step -
accuracy: 0.6491 - loss: 0.9036 - val accuracy: 0.8861 - val loss:
0.3376\n'',
      "Epoch 2/10\n",
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 -\u001b[0m\u001b[37m\u001b[0m \u001b[1m1250s\u001b[0m 7s/step -
accuracy: 0.8932 - loss: 0.3079 - val accuracy: 0.9052 - val loss:
0.2685\n",
      "Epoch 3/10\n",
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accuracy: 0.9227 - loss: 0.2287 - val accuracy: 0.9236 - val loss:
0.2256\n",
      "Epoch 4/10\n",
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  -\u001b[0m\u001b[37m\u001b[0m \u001b[1m1152s\u001b[0m 7s/step -
accuracy: 0.9237 - loss: 0.2191 - val accuracy: 0.9236 - val loss:
0.2194\n",
      "Epoch 5/10\n",
      "\u001b[1m171/171\u001b[0m \u001b[32m-
  -\u001b[0m\u001b[37m\u001b[0m\u001b[1m1154s\u001b[0m 7s/step -
accuracy: 0.9375 - loss: 0.1783 - val accuracy: 0.9339 - val loss:
0.1999\n",
      "Epoch 6/10\n",
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--\u001b[0m\u001b[37m\u001b[0m \u001b[1m1219s\u001b[0m 7s/step -
accuracy: 0.9472 - loss: 0.1585 - val accuracy: 0.9309 - val loss:
0.2054\n",
      "Epoch 7/10\n",
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--\u001b[0m\u001b[37m\u001b[0m\u001b[1m1159s\u001b[0m 7s/step -
```

```
accuracy: 0.9434 - loss: 0.1664 - val accuracy: 0.9317 - val loss:
0.2041\n",
      "Epoch 8/10\n",
      "\u001b[1m171/171\u001b[0m \u001b[32m-
  -\u001b[0m\u001b[37m\u001b[0m \u001b[1m1207s\u001b[0m 7s/step -
accuracy: 0.9447 - loss: 0.1471 - val accuracy: 0.9412 - val loss:
0.1683\n",
      "Epoch 9/10\n",
      "\u001b[1m171/171\u001b[0m \u001b[32m-
  -\u001b[0m\u001b[37m\u001b[0m \u001b[1m1243s\u001b[0m 7s/step -
accuracy: 0.9522 - loss: 0.1379 - val accuracy: 0.9390 - val loss:
0.1741\n'',
      "Epoch 10/10\n",
      "\u001b[1m171/171\u001b[0m \u001b[32m-
---\u001b[0m\u001b[37m\u001b[0m \u001b[1m1134s\u001b[0m 7s/step -
accuracy: 0.9558 - loss: 0.1311 - val accuracy: 0.9493 - val loss:
0.1546\n"
     ]
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   ],
   "source": [
    "history = model.fit(\n",
         train generator, \n",
         validation data=val generator, \n",
         epochs=epochs\n",
    ")\n"
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      "WARNING: absl: You are saving your model as an HDF5 file via
`model.save()` or `keras.saving.save model(model)`. This file format is
considered legacy. We recommend using instead the native Keras format,
e.g. `model.save('my model.keras')` or `keras.saving.save model(model,
'my model.keras')`. \n"
     ]
    }
   ],
   "source": [
   "os.makedirs(\"model\", exist ok=True)\n",
   "model.save(\"model/poultry model.h5\")\n"
   ]
  },
```

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                   -\u001b[0m \u001b[1m3:39\u001b[0m 6s/step"
     ]
    }
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   "source": [
    "val generator.reset()\n",
    "preds = model.predict(val generator) \n",
    "y pred = np.argmax(preds, axis=1)\n",
    "y true = val generator.classes\n",
    "\n",
    "print(\"Classification Report:\\n\", classification report(y true,
y pred, target names=list(val generator.class indices.keys()))) \n",
    "print(\"Confusion Matrix:\\n\", confusion matrix(y true, y pred))\n"
   ]
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   "source": [
    "model = load model(\"model/poultry model.h5\")\n",
    "img path = \"static/uploaded/image1.jpg\"  # Replace with actual test
image path\n",
    "img = cv2.imread(img path) \n",
    "img = cv2.resize(img, (img width, img height)) \n",
    "img = img / 255.0\n",
    "img = np.expand dims(img, axis=0)\n",
    "\n",
    "prediction = model.predict(img)\n",
    "predicted class = np.argmax(prediction)\n",
    "\n",
    "class labels = list(train generator.class indices.keys()) \n",
    "print(\"Predicted Class:\", class labels[predicted class])\n"
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