

Topic	Model View Controller	
Class Description	Students learn to create a prediction model which takes data from an API and makes predictions.	
Class	C125	
Class time	45 mins	
Goal	 Create an API with the post request. Create a classifier model. Send data through the API to the classifier to make predictions. 	
Resources Required	 Teacher Resources Postman Software Laptop with internet connectivity Earphones with mic Notebook and pen Student Resources Postman Software Laptop with internet connectivity Earphones with mic Notebook and pen 	
Class structure	Warm Up Teacher-led Activity Student-led Activity Wrap up	5 mins 15 min 15 min 5 min

CONTEXT

• Explore the technique of sending data to the classifier through an API

Class Steps	Teacher Action	Student Action
Step 1: Warm Up (5 mins)	Hi <student name="">. How are you doing today? Wasn't last class fun? Can you tell me</student>	ESR: We learned about the flask framework.

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	what were the new things that you learned?	We use it to create an API. We tested our API on the postman software.		
	Perfect!! So we got to know how we can create an API. And we also saw how we can add data to an API using a POST request. In the last few classes we have created multiple prediction models and have provided them data in various ways. Today we are going to do one more such thing where we pass the data through the API that we create and then see how well our prediction model performs. Sounds exciting?	ESR: Yes!!		
	Let's get started then			
	Teacher Initiates Screen Share			
CHALLENGE • Create the classifier model which takes data from the API				
Step 2: Teacher-led Activity (15 min)	<teacher code="" editor="" opens="" the="" vs=""> Can you tell me what are the steps we follow while creating a prediction model?</teacher>	ESR: 1.We first import all the dependencies/ libraries. 2. Then we get the data to train and test the model. 3. Then we split the data to train and test the model. 4. Then we scale the data to make it of equal size. 5. Then we fit the data into		



	the logistic regression model. 6.Then we make predictions.
Yes! Alright so let's start with our prediction model to which we'll be providing the data through the API that we create.	-
Yes! <teacher a="" all="" and="" are="" classifier.py="" codes="" creates="" file="" for="" import="" libraries="" model.="" needed="" that="" the="" to=""> Code: import numpy as np import pandas as pd from sklearn.datasets import fetch_openml from sklearn.model_selection import train_test_split from sklearn.linear_model import LogisticRegression from PIL import Image import PIL.ImageOps</teacher>	ESR: To Import all the libraries.

```
import numpy as np
import pandas as pd
from sklearn.datasets import fetch_openml
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
from PIL import Image
import PIL.ImageOps
```



Now we have all the libraries imported, now it's time to import the data that we'll be using to train and test the model.

A class before we learned to take the data from the data set provided by the machine learning dataset provided by the sklearn.

So we are going to use the digits image data that we used before in one of our previous classes.

<Teacher codes to import the data.>
Code:

X, y = fetch_openml('mnist_784', version=1, return_X_y=True)

Here we are using the fetch_openml function to get data, the name of the data set is 'mnist_784, version =1, and getting the x and y values of it.

The student helps the teacher with the code.

X, y = fetch_openml('mnist_784', version=1, return_X_y=True)

What's the next step?

<Teacher codes to split the data to train and test the model.>
Code:

X_train, X_test, y_train, y_test = train_test_split(X, y, random_state=9, train_size=7500, test_size=2500)

What is the ideal training and testing

ESR:

To split the data to train and test the model.

ESR:

75% for training ,25% for testing.

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size of the data? Yes. X train, X test, y train, y test = train_test_split(X, y, random_state=9, train size=7500, test size=2500) Now we need to scale the data to The student helps the make sure that the data points in X teacher with the code. and Y are equal so we'll divide them using 255 which is the maximum pixel of the image. <Teacher codes to scale the data.> Code: X train scaled = X train/255.0 X test scaled = X test/255.0 X train scaled = X train/255.0 test scaled = X test/255.0 After scaling the data we have to fit it The student helps teacher inside our model so that it can give with the code. output with maximum accuracy. <Teacher codes to fit the data in the logistic regression model.> Code: clf = LogisticRegression(solver='saga', multi_class='multinomial').fit(X_trai n_scaled, y_train)



clf = LogisticRegression(solver='saga', multi_class='multinomial').fit(X_train_scaled, y_train)

Now, we have our classifier ready. Using this classifier, if we have an image, can we predict the digit mentioned on the image?

If we remember what we did earlier, we were using CV2 and using CV2, we were using our device's camera and capturing each frame. Now, each frame was an image where we were doing some processing and then predicting the value from it.

Let's create a function to do that.

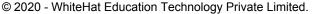
We'll call it get_prediction which will take the image as the parameter and make a prediction.

This function will take the image and convert it into a scalar quantity and then make it grey so that the colors don't affect the prediction.

And then resize it into 28 by 28 scales. Then using the percentile function get the minimum pixel and and then using the clip function give each image a number. And then get the maximum pixel and make an array of this. And then create a test sample of it and make predictions based on that sample. And then finally return the test prediction.

ESR:

Yes.





```
Code:
def get prediction(image):
  im pil = Image.open(image)
  image bw = im pil.convert('L')
  image bw resized =
image_bw.resize((28,28),
Image.ANTIALIAS)
  pixel filter = 20
  min pixel =
np.percentile(image bw resized,
pixel filter)
image bw resized inverted scale
d =
np.clip(image bw resized-min pix
el, 0, 255)
  max pixel =
np.max(image_bw_resized)
image bw resized inverted scale
d =
np.asarray(image bw resized inve
rted scaled)/max pixel
  test sample =
np.array(image_bw_resized_invert
ed scaled).reshape(1,784)
  test pred =
clf.predict(test sample)
  return test pred[0]
```

```
def get_prediction(image):
    im_pil = Image.open(image)
    image_bw = im_pil.convert('L')
    image_bw_resized = image_bw.resize((28,28), Image.ANTIALIAS)
    pixel_filter = 20
    min_pixel = np.percentile(image_bw_resized, pixel_filter)
```

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```
image bw resized inverted scaled = np.clip(image bw resized-min_pixel,
0, 255)
   max pixel = np.max(image bw resized)
   image bw resized inverted scaled =
np.asarray(image bw resized inverted scaled)/max pixel
   test sample = np.array(image bw resized inverted scaled).reshape(1,784)
   test_pred = clf.predict(test_sample)
   return test_pred[0]
                                                        ESR:
                   Now that our classifier is complete,
                                                        We can create an API which
                   how can we pass an image into this
                   function that we just created?
                                                        accepts an image.
                                                        ESR:
                   Can you try writing the API and use
                                                        Yes.
                   the classifier function in it?
                   Let's get started then
                           Teacher Stops Screen Share
                   Now it's your turn. Please share your
                   screen with me.

    Ask Student to press ESC key to come back to panel

            Guide Student to start Screen Share

    Teacher gets into Fullscreen

                                   ACTIVITY

    Create an API with POST request

    Send data through the API and test the classifier model
```



Step 3: Student-Led Activity (15 min)	<teacher a="" and="" classifier.py="" code="" create="" editor="" file="" function.="" get_prediction="" helps="" open="" student="" the="" to="" write=""></teacher>	Student opens the code editor, creates a classifier.py and writes code for the get_prediction function.
	Alright now we'll start creating our API. First we'll import Flask, requests to make request on the API and jsonify to convert data into json.	Student codes to import Flask, requests and jsonify from flask.
	<teacher ,requests="" and="" flask="" flask.="" from="" helps="" import="" jsonify="" student="" to=""> Code: from flask import Flask, jsonify, request</teacher>	dingfor
from flask impo	ort Flask, jsonify, request	
	As we know that every python program is a module and can be used	Students code to import the get_prediction function.

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code.> Code: from classifier import get_prediction from classifier import get prediction We'll now give the name of the app to The student codes to set the name of the app in the flask the flask constructor so that it'll know where to look for the dependent files. constructor. Code: app = Flask(__name__ app = Flask(name Now we'll start writing our route. Student codes to write the We need a post request to send the route with a post request. image to the prediction model and our route name would be 'predict-digit'. Code: @app.route("/predict-digit", methods=["POST"])

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```
@app.route("/predict-digit", methods=["POST"])
```

Inside this route we'll write a function called predict_data().

We'll use the request function to get the files from the digit key.
We'll also use the get_prediction

function here and return the prediction in the json format with the status code of 200.

Code:

```
def predict_data():
    image = request.files.get("digit")
    prediction =
    get_prediction(image)
    return jsonify({
        "prediction": prediction
    }), 200
```

The student codes to write the predict_data function.

```
def predict_data():
    image = request.files.get("digit")
    prediction = get_prediction(image)
    return jsonify({
        "prediction": prediction
    }), 200
```



Then we'll use the' if __name__ == "main" ' block to prevent (certain) code from being run when the module is imported.

Student codes to write if__name__==main block.

<Teacher helps student with the code.>

Code:

if __name__ == "__main__":
 app.run(debug=True)

We write debug=True so that the server will update every time you save the code.

```
if __name__ == "__main__":
    app.run(debug=True)
```

Now we'll run the code, copy the port on which the code is running and test the API in the postman.

<Teacher asks student to open the postman.>

Now make a post request on the route that we created which would be "http://127.0.0.1:5000/predict-digit".

We have to make the request in the "form-data" as we'll be giving image files to the API.

In the place of "key" we have to write "digit" and type as file. As its value

Student opens the postman and tests the API and the prediction model.

The student uses the images given in student activity 1 to test the prediction.

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we'll be using the digit images files. And then make a request. To see the prediction. cop/API\$ python3 app.py /home/ashura/.local/lib/python3.8/site-packages/sklearn/linear_model/_sag.py:329: ConvergenceWarn ched which means the coef_ did not converge warnings.warn("The max_iter was reached which means "
* Serving Flask app "app" (lazy loading) * Environment: production Use a production WSGI server instead. * Debug mode: off * Running on http://127.0.0.1:5000/ (Press CTRL+C to quit) Untitled Request http://127.0.0.1:5000/predict-digit ✓ dig **Teacher Guides Student to Stop Screen Share** FEEDBACK Appreciate the student for their efforts Identify 2 strengths and 1 area of progress for the student Step 4: How did you find today's class? Student talks about his/her Wrap-Up experience of today's class. (5 min)

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So now we know how we can send data through an API to test the prediction model. Before we end the class, can you tell me why we created a classification model in a separate file instead of doing it in the api file itself?	ESR: The Classifier takes time to learn from the training data that we provide. If we build it in the API itself, then it will take a lot of time to give the response.
Also, why did we make a separate function to process the image and predict the outcome, instead of doing it in the API?	ESR: Flask APIs, to optimize the performance speed, caches the APIs so that whenever an API call is made, it can return the data much faster. If we did this in the API itself, it will return the same output for different images that we send to the server.
What we did today is known as the MVC Architecture. MVC stands for Model View Controller .There are also other types of architectures like MVVM,MVP, MVA etc. Keeping in mind that the API caches the data so that it can give a response faster, it is ideal to keep your processing code separately from your API.	

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	Model is generally referred to the Database, if any. Here, we are not using any Databases.	
	View is referred to the API, which accepts a request and returns a response.	
	Controller is the part that does all the heavy work, i.e, data processing, building classifier, etc.	* ids
	Hence, it is known as the MVC Architecture!	O tolk
	In our next class we'll do something much more amazing. Are you up for it?	dins
	Alright then, I'll see you in next class.	
	Teacher Clicks × End Class	
Additional Activities	Encourage the student to write reflection notes in their reflection journal using markdown.	The student uses the markdown editor to write her/his reflection in a reflection journal.
	 What happened today? Describe what happened Code I wrote How did I feel after the class? 	

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 What have I learned about programming and developing games? What aspects of the class helped me? What did I find 	
difficult?	

Activity	Activity Name	Links
Teacher Activity 1	Reference code	https://github.com/whitehatjr/C125/tr ee/master/API
Student Activity 1	Images for testing	https://github.com/whitehatjr/C125/tr ee/master/API/images

