## **Assignment - 6**

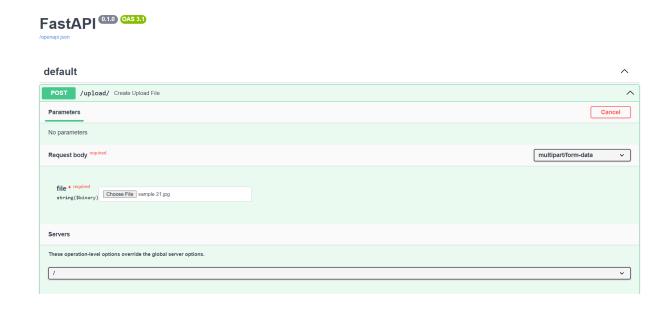
## Build a FastAPI for MNIST digit prediction

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### Introduction:

A contemporary, quick framework for creating Python APIs is called FastAPI.Its asynchronous programming and high speed architecture enable it to manage several concurrent connections with ease. Additionally, we can run our apps on the local server via its Swagger UI. Here, we attempt to put up an API for the MNIST data and do the tests on it. Since the second model from the last assignment produced one of the greatest test accuracies, we employ it.

We apply the functions as described in the question document. On running the commands, and importing uvicorn and loading swagger UI which is a web server for running ASGI applications.: we get this interface



Initially, we provide the API with a 28x28 image from MNIST, as we have not yet developed the code for using arbitrary image sizes. The output is shown in the next figure.

To use arbitrary images from MS Paint, we resize them using `image.resize((28,28))` to get images of normal size. We ensure that the background is black, just like the MNIST dataset, and draw accordingly. We conduct 10 experiments and obtain the following results:

- For 8, the output is correct.
- For 5, the output is correct.
- For 6, the output is correct.
- For 9, the model gets it right.
- For 3, the output is correct.
- For 0, the output is correct.
- For 5, the output is correct.
- For 1, the output is correct.
- For 2, the output is correct.
- For 7, the output is correct.

### \*\*Conclusion\*\*

We have explored various MLflow commands, including `start run`, `autolog`, `mlflow.log params`, and `mlflow.log metrics`, as well as nested runs. We have observed the various use cases and obtained a comparison between various models used for MNIST. We find that Model 8 has slightly better overall test accuracy. We also gain insight into the variability of various metrics across various experiments, demonstrating the usefulness of MLflow in streamlining and quickening model analysis.





