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Step 3:

1) Clone repo

2) Navigate to Anomaly Detection/inference

```
S C:\Users\darre\OneDrive\Desktop\School\grad\_Semester5\Data Eng\project\AI-for-Astronomy> ls
    Directory: C:\Users\darre\OneDrive\Desktop\School\grad\_Semester5\Data Eng\project\AI-for-Astronomy
                                                 Length Name
                        LastWriteTime
                 7/22/2025
                              8:41 PM
                                                         .ipynb_checkpoints
                 7/22/2025
                               8:35 PM
                              8:35 PM
                 7/22/2025
                                                        code
                 7/22/2025
                              8:41 PM
                                                  papers
256 .gitignore
11558 LICENSE
                 7/22/2025
                              8:35 PM
                 7/22/2025
                                                   9029 README.md
Directory: C:\Users\darre\OneDrive\Desktop\School\grad\_Semester5\Data Eng\project\AI-for-Astronomy\code
                        LastWriteTime
                                                 Length Name
                 7/22/2025 8:35 PM
                                                        Anomaly Detection
PS C:\Users\darre\OneDrive\Desktop\School\grad\_Semester5\Data Eng\project\AI-for-Astronomy\code> cd a*
PS C:\Users\darre\OneDrive\Desktop\School\grad\_Semester5\Data Eng\project\AI-for-Astronomy\code\Anomaly Detection> ls
    Directory: C:\Users\darre\OneDrive\Desktop\School\grad\_Semester5\Data Eng\project\AI-for-Astronomy\code\Anomaly Detection
                        LastWriteTime
                                                 Length Name
                 7/22/2025
                                                        blocks
                              8:35 PM
                                                         Fine_Tune_Model
                 7/22/2025
                              8:40 PM
                                                        Inference
                                                        Plots
                 7/22/2025
                                                2405631 Astronomy_Overview.pptx
5407 NormalCell.py
2716 Plot_Redshift.py
                 7/22/2025
                 7/22/2025
                              8:35 PM
                 7/22/2025
                                                  11130 README.md
PS C:\Users\darre\OneDrive\Desktop\School\grad\_Semester5\Data Eng\project\AI-for-Astronomy\code\Anomaly Detection> cd In*
PS C:\Users\darre\OneDrive\Desktop\School\grad\_Semester5\Data Eng\project\AI-for-Astronomy\code\Anomaly Detection\Inference> ls
    Directory: C:\Users\darre\OneDrive\Desktop\School\grad\_Semester5\Data Eng\project\AI-for-Astronomy\code\Anomaly Detection\Inference
```

3) Update file paths in inference.py to match local env

Wasn't needed

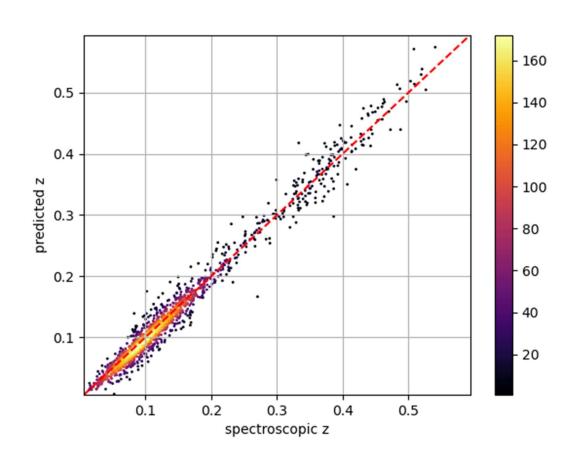
4) Run inference.py script and document execution time

The astronomy inference was successfully executed on a CPU-only machine using a batch size of 512 across 3 batches. The total CPU execution time was approximately 12.70 seconds, with an average per-batch time of 4.23 seconds. Peak CPU memory usage was ~25.3 GB, and throughput was measured at ~16.2 Mbps.

- 5) Capture output results
 - a. Results.json

total cpu time (second) 12.698254299999993
total gpu time (second) 0
execution time per batch (second) 4.232751433333331
cpu memory (MB) 25336.653888
gpu memory (MB) 0
throughput(bps) 16198484.857875315
batch size 512
number of batches 3
device "cpu"
MAE 0.01251969638102928
MSE 0.0002972779279695162
Bias 0.0020244880761514164
Precision 0.011360458992421626
R2 0.9746744148433208

b. inference.png



- 6) Provide a brief analysis of the inference performance on your system
 The inference ran on a CPU-only system with a batch size of 512. It processed 3
 batches in about 12.7 seconds total, averaging just over 4 seconds per batch.
 The memory usage was around 25 GB, which is a bit high, but expected for a
 large model. The predictions were very close to the actual values, with a low
 error rate and a high R² score of 0.97, meaning the model was very accurate.
 Overall, it performed well even without a GPU.
- 7) Document any errors encountered

The only errors encountered were version errors for certain packages. Through trial and error, we were able to sort those out (pip uninstall ... pip install...==x.x.xx)