## S3LLM Docker Tutorial

To follow this tutorial, ensure you have the latest versions of Docker and Nvidia CUDA installed.

1. Download the Docker image using this link: https://mega.nz/folder/R2xTCLjC#dUrUkopEGuCT-OC4iejhVw.

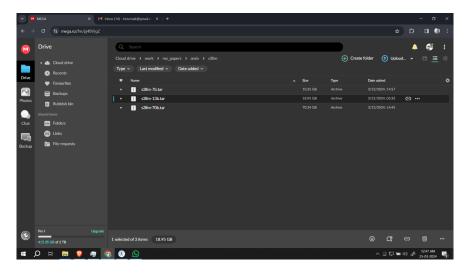


Figure 1: Download S3LLM docker images

2. Once you've downloaded the LLaMA-2 specific model Docker image, load it into Docker using the following command:

```
docker load < [file_name]</pre>
```

```
docker load < s3llm-13b.tar</pre>
```

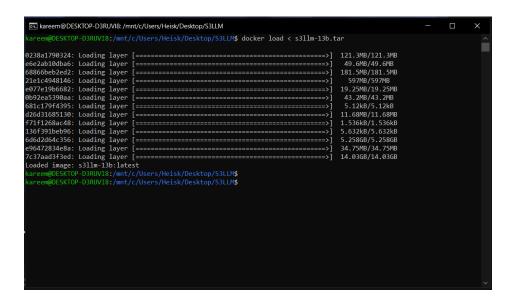


Figure 2: Load S3LLM Docker images

3. Verify if the image has been loaded properly by running:

docker image list

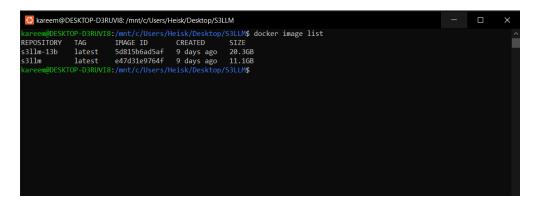


Figure 3: GUI Version of Docker

4. To run the loaded image, use the following command:

```
docker run --gpus all -it [image_name]
docker run --gpus all -it s3llm-13b
```

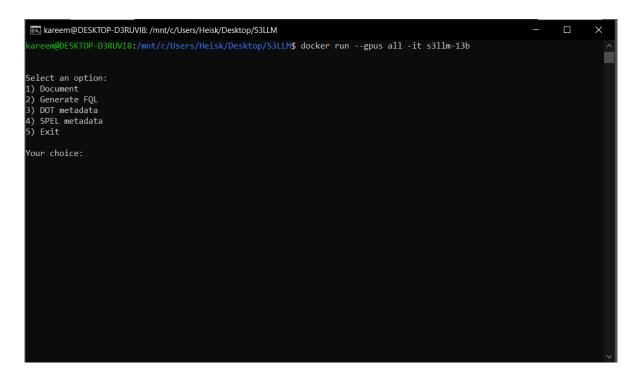


Figure 4: Running Docker Image

5. In Figure 5, I have selected the first option to perform document analysis. Figure 6 displays the generated result.

Figure 5: Document Analysis

```
kareem@DESKTOP-D3RUVI8: /mnt/c/Users/Heisk/Desktop/S3LLM
llama_print_timings:
                                                   load time =
                                                                            58667.15 ms
llama_print_timings: sample time =
llama_print_timings: prompt eval time =
                                                                               157.54 ms /
                                                                                                                                     0.62 ms per token, 1625.03 tokens per second)
                                                                                                        256 runs (
288 tokens (
                                                                           58666.69 ms /
82347.97 ms /
                                                                                                                                                                               4.91 tokens per second)
                                                                                                                                  203.70 ms per token,
llama_print_timings:
                                                                                                                                  322.93 ms per token,
                                                  eval time =
                                                                                                                                                                               3.10 tokens per second)
llama_print_timings:
                                                 total time =
                                                                          142160.84 ms /
                                                                                                         543 tokens
AI: The Crank-Nicholson method is applied in this model by solving the system of equations obtained from Equation (9.44
) using an implicit treatment of the time derivative. Specifically, at each time step t, the method updates the values o
    the variables in each layer using the following updates:
\ \begin{aligned} \frac{W_i^{n+1} - W_i^n}{\Delta t} &= \frac{1}{2} \left( \frac{F_i^n}{\Delta t} + \frac{F_i^{n-1}}{\Delta t} \right) \ , \\ \frac{h_i^{n+1} - h_i^n}{\Delta t} &= \frac{1}{2} \left( \frac{L_i^n}{\Delta t} + \frac{L_i^{n-1}}{\Delta t} \right) \ , \\ \frac{\phi_i^{n+1} - \phi_i^n}{\Delta t} &= \frac{1}{2} \left( \frac{R_i^n}{\Delta t} + \frac{Delta t} + \frac{Delta t} \right) \ , \\ \frac{\phi_i^{n+1} - \phi_i^n}{\Delta t} &= \frac{1}{2} \left( \frac{R_i^n}{\Delta t} + \frac{Delta t} + \frac{Delta t} \right) \ , \\ \frac{\phi_i^n}{\Delta t} \right\} \right\}
 (R_i^{n-1
Select an option:
1) Document
2) Generate FQL
3) DOT metadata
4) SPEL metadata
5) Exit
 Your choice:
```

Figure 6: Result