# Handbook

# **Setup Environment**

- 1. Make a fresh installation of Manjaro.
- 2. Download the repo from Github.
- 3. a) Download models from the github links, and put them in a folder, say /run/media/xxx/data/c491/c4-models-work/.

```
Name

- I ABL_a_asc_2x_lsct3sp_lsct3spva9r.zip

- I baseline.zip

- I DUAL_a_asc_0dancukmk7hnp_r45_C_trinorm_dsa3_va9r.zip

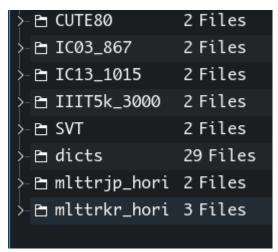
- I DUAL_a_asc_0dancukmk7hnp_r45pt_C_trinorm_dsa3_va9r_lsct3sp_2x.zip

- I DUAL_b_asc_0dancukmk7hnp_r45_C_trinorm_dsa3_va9r_lsct3sp_2x.zip

- I DUAL_b_asc_0dancukmk7hnp_r45pt_C_trinorm_dsa3_va9r_lsct3sp_2x.zip

- I DUAL_ch_asc_0dancukmk7hnp_r45_C_trinorm_dsa3_va9r_lsct3sp_2x.zip
```

- b) Run for i in \$(ls); do unzip \$i; done; to uncompress them
- 4. Download datasets from the github links and uncompress it to /run/media/xxx/data/xxx/ssddata/



5. Uncompress the Athena.zip file to /run/media/xxx/data/xxx/ssddata/

- 6. Install dependencies
  - a) Install packages on paclist.txt in the git repo
  - b) Install pip packages on the piplist.txt in the git repo
  - c) Install pylcs-laser.zip

pylcs modded by a Github user, adds support to return the exact LCS

Source: <a href="https://github.com/Meteorix/pylcs/issues/4">https://github.com/Meteorix/pylcs/issues/4</a>,

7. Change the paths in neko\_sdk/root.py to the data and models uncompressed. If you do NOT want intermediate results, return None for find\_export\_root.

```
sdanmk7_routine_cfg_va9.py × & eval_configs.py × & dual_lsct32.py × & dual_no_lsct_32.py × & test.py × & root.py ×
import ...

def find_data_root():
    return "/run/media/xxxx/testing/home/xxxx/c491/ssddata/"

def find_model_root():
    return "/run/media/xxxx/testing/home/xxxx/c491/models-release/"

def find_export_root():
    return "/run/media/xxxx/testing/home/xxxx/c491/results/"
```

All done! You can now go to the next steps to check out if our framework lives up to its name.

#### **Evaluation:**

Going  $\sim$ 100 FPS on a laptop single batched is what we reported in the paper, now let's go 280+ FPS on that laptop going multi-batched. We actually liked the OSOCR way of testing on laptops so we also perform evaluations on laptops consuming around 230w (120w GPU+65W CPU + a little bit extra). This gives an insight into how the model may behave when deployed on a mobile platform that fits into vehicles or other power-consumption-sensitive computing platforms.

# **Reproducing Ablative Study**

Base model: Run test\_rej.py under DUAL\_a\_asc\_Odancukmk7hnp\_r45\_C\_trinorm\_dsa3

CIL only: Run test\_rej.py under DUAL\_a\_asc\_Odancukmk7hnp\_r45\_C\_trinorm\_dsa3\_va9r

```
JAP_lang starts
0.0831947298042790256 4009 FPS: 313.0155165737643
0.08319472980427902774 4009

I Dase_chs_close_set_benchmark]test_accr
Accuracy: 0.391868, AR: 0.608202, CER: 0.391798, WER: 0.608132
(0.61)
(0.61)
(0.61)
JAP_lang ends
9
JAP_lang starts
0.0833062264504626555 4009 FPS: 294.4444413780988
0.083306226454647631416199159044009
[Dase_chs_close_set_benchmark]test_accr
KACR: 0.814341, URCL: 0.816914, UPRE 0.974861, F 0.888926
[Dase_chs_close_set_benchmark]test_accr
KACR: 0.814341, URCL: 0.816914, UPRE 0.974861, F 0.888926
0.08336624728435490666 4009 FPS: 273.6522538992436
0.0837574275144931440910
[Dase_chs_close_set_benchmark]test_accr
KACR: 0.873103, URCL: 0.524130, UPRE 0.846782, F 0.647487
[Dase_chs_close_set_benchmark]test_accr
KACR: 0.673103, URCL: 0.524130, UPRE 0.846782, F 0.647487
[Dase_chs_close_set_benchmark]test_accr
KACR: 0.673103, URCL: 0.524130, UPRE 0.846782, F 0.647487
[Dase_chs_close_set_benchmark]test_accr
KACR: 0.673103, URCL: 0.524130, UPRE 0.846782, F 0.647487
[Dase_chs_close_set_benchmark]test_accr
KACR: 0.673103, URCL: 0.524130, UPRE 0.846782, F 0.647487
[Dase_chs_close_set_benchmark]test_accr
KACR: 0.673103, URCL: 0.524130, UPRE 0.846782, F 0.647487
[Dase_chs_close_set_benchmark]test_accr
KACR: 0.673103, URCL: 0.524130, UPRE 0.846782, F 0.647487
[Dase_chs_close_set_benchmark]test_accr
KACR: 0.673103, URCL: 0.524130, UPRE 0.846782, F 0.647487
[Dase_chs_close_set_benchmark]test_accr
KACR: 0.673103, URCL: 0.524130, UPRE 0.846782, F 0.647487
[Dase_chs_close_set_benchmark]test_accr
KACR: 0.673103, URCL: 0.524130, UPRE 0.846782, F 0.647487
[Dase_chs_close_set_benchmark]test_accr
KACR: 0.673103, URCL: 0.524130, UPRE 0.846782, F 0.647487
[Dase_chs_close_set_benchmark]test_accr
KACR: 0.673103, URCL: 0.524130, UPRE 0.846782, F 0.647487
[Dase_chs_close_set_benchmark]test_accr
```

I think it can run even faster, but nevermind.

GZSL OSR GOSR

### Reproducing Results On English, Japanese, and Korean

Regular model: Run test.py under ABL\_a\_asc\_2x\_lsct3sp\_lsct3spva9r

```
CUTE starts
0.010705261594719358 288 FPS: 93.41200970682262
0.014507010579109192 288
[base_chs_lsctsp_2x_va9r_close_set_benchmark]test_accr
Accuracy: 0.729167, AR: 0.870846, CER: 0.129154, WER: 0.270833
[base_chs_lsctsp_2x_va9r_close_set_benchmark]test_accr
Accuracy: 0.729167, AR: 0.870846, CER: 0.129154, WER: 0.270833
(0, \{\})
CUTE ends
IIIT5k starts
0.009958661953608195 3000 FPS: 100.41509639130614
0.009964030663172405 3000
[base_chs_lsctsp_2x_va9r_close_set_benchmark]test_accr
Accuracy: 0.832000, AR: 0.936669, CER: 0.063331, WER: 0.168000
Accuracy: 0.832000, AR: 0.936669, CER: 0.063331, WER: 0.168000
(0, \{\})
IIIT5k ends
JAP_lang starts
0.003196817587426429 4009 FPS: 312.8110918599649
0.0033238392037981375 4009
Accuracy: 0.415565, AR: 0.617967, CER: 0.382033, WER: 0.584435
[base_chs_lsctsp_2x_va9r_close_set_benchmark]test_accr
(0, \{\})
KR_lang starts
0.0037015587281636723 5171 FPS: 270.15645932925554
0.0037829781426979526 5171
Accuracy: 0.195320, AR: 0.438824, CER: 0.561176, WER: 0.804680
[base_chs_lsctsp_2x_va9r_close_set_benchmark]test_accr
Accuracy: 0.195320, AR: 0.438824, CER: 0.561176, WER: 0.804680
(0, \{\})
```

The FPSs are incoherent due to batch size.

Also note the running time includes prototype caching time, hence CUTE(288) has a lower FPS than IIIT5k(3000). In practical use, the FPS will be close to IIIT5k performance.

Large: Run test.py under DUAL\_a\_asc\_Odancukmk7hnp\_r45pt\_C\_trinorm\_dsa3\_va9r\_lsct3sp\_2x

```
ACCUPACY: 0.784722, AK: 0.894671, CEK: 0.105329, WEK: 0.215278
î
   0.015140331427256267 3000 FPS: 66.04875228819348
   [base_chs_close_set_benchmark]test_accr
   [base_chs_close_set_benchmark]test_accr
   Accuracy: 0.850667, AR: 0.945118, CER: 0.054882, WER: 0.149333
   IIIT5k ends
   0.005023633610075268 4009 FPS: 199.0591029557622
   0.005359705798671738 4009
   [base_chs_close_set_benchmark]test_accr
   Accuracy: 0.444749, AR: 0.649319, CER: 0.350681, WER: 0.555251
   (0, {})
JAP_lang ends
   99
   0.0051070246791452304 5171 FPS: 195.80872676874773
   0.005276509766144348 5171
   Accuracy: 0.221427, AR: 0.454455, CER: 0.545545, WER: 0.778573
   KR_lang ends
   Q Find ▶ Run ≔ TODO • Problems * Debug * Python Packages ☑ Termi
```

## **Reproducing Close-set Benchmarks**

Regular: run test.py under DUAL\_b\_asc\_Odancukmk7hnp\_r45\_C\_trinorm\_dsa3\_va9r\_lsct3sp\_2x

```
<u>/usr/lib/python3.10/site-packages/torch/utils/data/dataloade</u>r.py:487
                                                                                 SVT starts
                                                                                 0.010439500558136788 647 FPS: 95.79002313674641
                                                                                 0.010465028297035182 647
      CUTE starts
      0.012077465653419495 288 FPS: 82.79882789125298
==
      0.015858748720751867.288
      [base_mjst_close_set_benchmark]test_accr
      Accuracy: 0.795139, AR: 0.875235, CER: 0.124765, WER: 0.204861
                                                                                 (0, {})
SVT ends
      [base_mjst_close_set_benchmark]test_accr
      Accuracy: 0.795139, AR: 0.875235, CER: 0.124765, WER: 0.204861
      (0, \{\})
                                                                                 0.010770864552692558 867
      CUTE ends
      IIIT5k starts
      0.010423674662907919 3000 FPS: 95.93545772859218
                                                                                 Accuracy: 0.925029, AR: 0.967801, CER: 0.032199, WER: 0.074971
      0.010429205020268758 3000
      [base_mjst_close_set_benchmark]test_accr
                                                                                 IC13 starts
      Accuracy: 0.892333, AR: 0.955465, CER: 0.044535, WER: 0.107667
                                                                                 0.010621431425874456 1015 FPS: 94.14926857823879
      IIIT5k ends
      SVT starts
      0.010439500558136788 647 FPS: 95.79002313674641
                                                                                 Accuracy: 0.903448, AR: 0.967724, CER: 0.032276, WER: 0.096552
```

Large: run test.py under DUAL\_b\_asc\_Odancukmk7hnp\_r45pt\_C\_trinorm\_dsa3\_va9r\_lsct3sp\_2x

```
/usr/lib/python3.10/site-packages/torch/utils/data/dataloader_____sends
                                                                      SVT starts
                                                                     0.015122226069478384 647 FPS: 66.12783034756558
99
                                                                     0.015148045290750919 647
CUTE starts
                                                                     Thase mist close set benchmarkltest accr
0.01571953876150979 288 FPS: 63.61509807454137
0.019835591316223145 288
                                                                     Thase mist close set benchmarkItest accr
[base_mjst_close_set_benchmark]test_accr
                                                                     (0, \{\})
Accuracy: 0.815972, AR: 0.900313, CER: 0.099687, WER: 0.184028
                                                                     (0, \{\})
[base_mjst_close_set_benchmark]test_accr
                                                                      SVT ends
Accuracy: 0.815972, AR: 0.900313, CER: 0.099687, WER: 0.184028 IC03 starts
                                                                     0.015496390363611839 867 FPS: 64.53115703307076
                                                                     0.015515541397851506 867
(0, \{\})
CUTE ends
IIIT5k starts
                                                                     Accuracy: 0.922722, AR: 0.967801, CER: 0.032199, WER: 0.077278
0.014877764304478963 3000 FPS: 67.2143999282842
0.014882989486058554 3000
[base_mjst_close_set_benchmark]test_accr
Accuracy: 0.915333, AR: 0.965486, CER: 0.034514, WER: 0.084667
                                                                     IC13 starts
[base_mjst_close_set_benchmark]test_accr
                                                                     0.017570234636955073 1015
Accuracy: 0.915333, AR: 0.965486, CER: 0.034514, WER: 0.084667
                                                                     Accuracy: 0.903448, AR: 0.969950, CER: 0.030050, WER: 0.096552
(0, \{\})
                                                                     [base_mjst_close_set_benchmark]test_accr
                                                                     Accuracy: 0.903448, AR: 0.969950, CER: 0.030050, WER: 0.096552
IIIT5k ends
                                                                     (0, \{\})
SVT starts
0.015122226069478384 647 FPS: 66.12783034756558
                                                                      IC13 ends
```

### **Reproducing Zero-shot Character Benchmarks**

Run test.py under DUAL\_ch\_asc\_Odancukmk7hnp\_r45\_C\_trinorm\_dsa3\_va9r\_lsct3sp\_2x

Note the datasets are not uploaded for review due to the size limitation.

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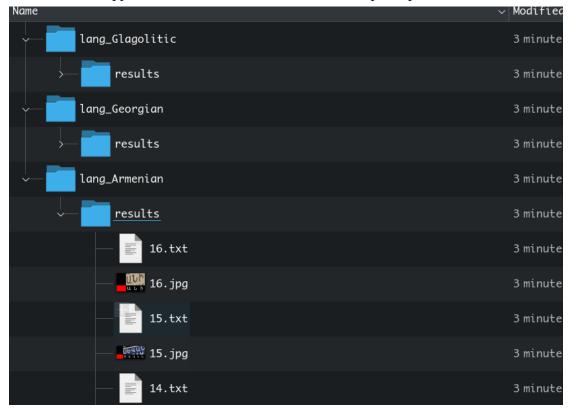
$7.500

$7.500
```

Speed drops after the card overheats, poor little card.

### **Recognition of some Greek Family Languages**

Run test\_athena.py under DUAL\_a\_asc\_Odancukmk7hnp\_r45pt\_C\_trinorm\_dsa3\_va9r\_lsct3sp\_2x



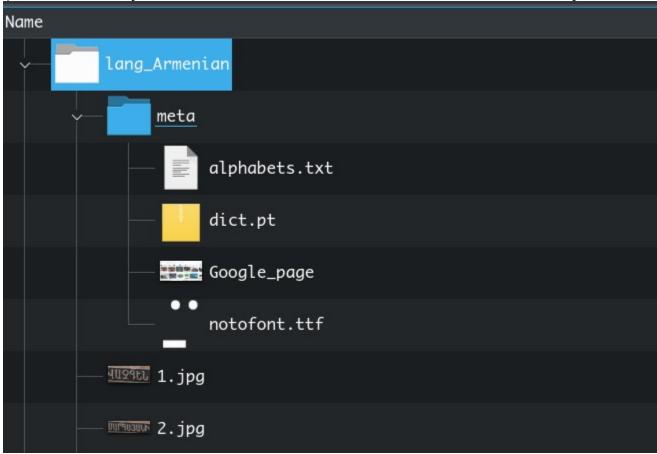
If you have your own language in mind, say xxxx language, you can test it with the following steps:

- a) add a folder named lang\_xxxx in the Athena folder (the lang\_xxxx part is mandatory for the testing script to pick it up)
- b) add the images in question to the folder
- c) Add a folder name meta (again it has to be named as `meta')
- d) Download Noto sans regular font in question to the meta folder and name it as `notofont.ttf'
- e) add the alphabets file, each character per line

if the character has more than one possible shape, write all of them in one line



f) Done! Don't worry about the dict file, the framework will build it from notofont and alphabets



# **System information**

Class	Description
system	Computer
bus	Motherboard
memory	24GiB System memory
processor	Intel(R) Core(TM) i5-9400 CPU @ 2.90GHz
bridge	8th Gen Core Processor Host Bridge/DRAM Registers
bridge	6th-10th Gen Core Processor PCIe Controller (x16)
display	TU106M [GeForce RTX 2070 Mobile]
multimedia	TU106 High Definition Audio Controller
display	CoffeeLake-S GT2 [UHD Graphics 630]