Study LSTM

DSC 180B - Group 3 - 2023/02/03

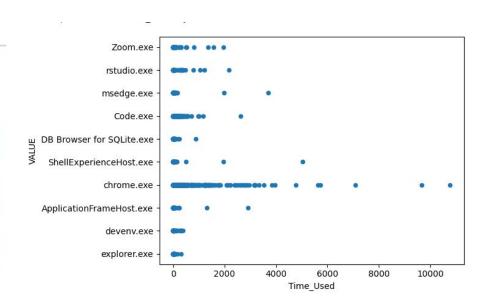


	MEASUREMENT_TIME	ID_INPUT	VALUE	PRIVATE_DATA	Time_Used
640	2023-01-17 06:01:02.341	4	chrome.exe	0	10768.311
1733	2023-01-23 00:42:20.235	4	chrome.exe	0	9673.942
796	2023-01-18 07:44:57.400	4	chrome.exe	0	7089.715
1630	2023-01-21 06:18:44.711	4	chrome.exe	0	5721.912
1653	2023-01-22 00:03:13.554	4	chrome.exe	0	5625.497
2575	2023-01-27 03:29:11.800	4	ShellExperienceHost.exe	0	5040.236
1642	2023-01-21 17:58:49.178	4	chrome.exe	0	4782.524
2288	2023-01-25 19:25:15.942	4	chrome.exe	0	3959.602
1680	2023-01-22 18:05:45.815	4	chrome.exe	0	3832.000
307	2023-01-16 02:18:08.317	4	msedge.exe	0	3698.763
1269	2023-01-20 05:16:54.972	4	chrome.exe	0	3520.850
123	2023-01-15 21:36:30.521	4	chrome.exe	0	3348.958
1003	2023-01-19 05:18:25.091	4	chrome.exe	0	3197.018
2528	2023-01-27 02:05:43.849	4	chrome.exe	0	3183.326
494	2023-01-16 21:36:58.888	4	chrome.exe	0	3160.566
1646	2023-01-21 22:22:01.224	4	chrome.exe	0	3137.300
2270	2023-01-25 18:08:37.536	4	chrome.exe	0	2965.804

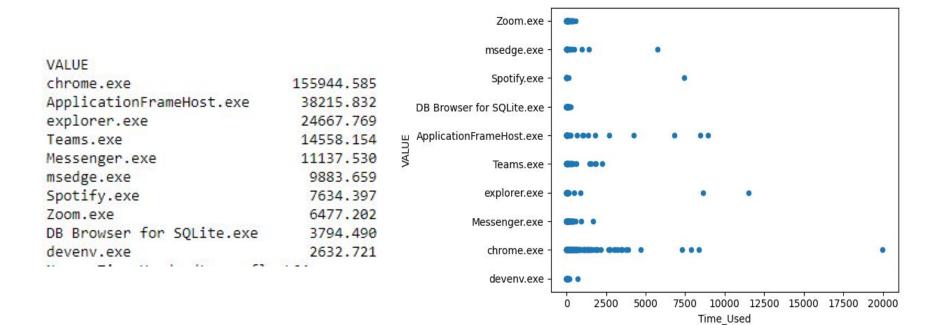
	MEASUREMENT_TIME	ID_INPUT	VALUE	PRIVATE_DATA	Time_Used
1263	2023-01-24 21:00:22.469	4	chrome.exe	0	19982.078
1005	2023-01-23 18:55:05.496	4	explorer.exe	0	11530.199
1727	2023-01-31 00:29:09.836	4	ApplicationFrameHost.exe	0	8935.665
1329	2023-01-26 00:37:02.316	4	explorer.exe	0	8611.865
1827	2023-02-01 00:48:18.343	4	ApplicationFrameHost.exe	0	8448.389
842	2023-01-22 15:39:14.564	4	chrome.exe	0	8370.358
1731	2023-01-31 03:17:15.001	4	chrome.exe	0	7864.547
1535	2023-01-27 03:34:56.038	4	Spotify.exe	0	7438.509
886	2023-01-23 01:42:40.339	4	chrome.exe	0	7291.960
1465	2023-01-27 01:03:21.880	4	ApplicationFrameHost.exe	0	6801.482
298	2023-01-20 00:56:02.975	4	msedge.exe	0	5761.546
858	2023-01-22 18:07:26.948	4	chrome.exe	0	4703.125
1705	2023-01-27 22:16:15.424	4	ApplicationFrameHost.exe	0	4254.468
1753	2023-01-31 06:23:40.426	4	chrome.exe	0	3870.576
870	2023-01-22 20:47:51.901	4	chrome.exe	0	3757.708
1400	2023-01-26 03:49:42.756	4	chrome.exe	0	3493.871
1181	2023-01-24 17:09:44.825	4	chrome.exe	0	3493.001
1414	2023-01-26 05:15:48.820	4	chrome.exe	0	3452.686
1434	2023-01-26 23:44:12.833	4	chrome.exe	0	3238.186

Data Preparation (User 1)

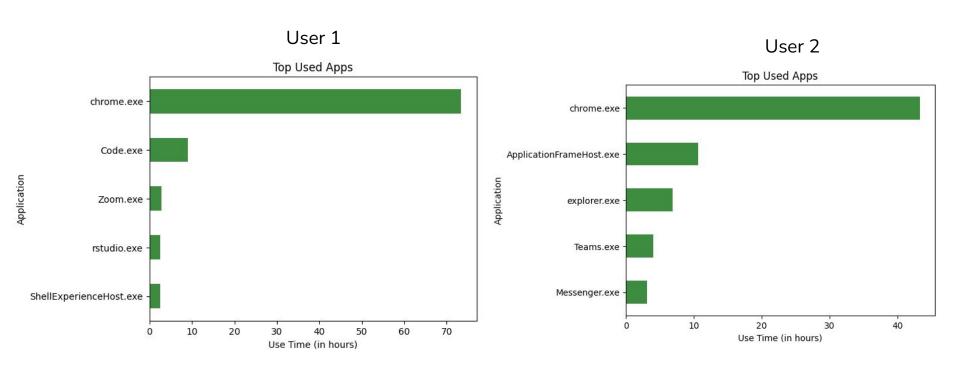
VALUE	
chrome.exe	264347.415
Code.exe	32619.039
Zoom.exe	10050.404
rstudio.exe	9281.864
ShellExperienceHost.exe	9026.440
msedge.exe	6833.118
ApplicationFrameHost.exe	5519.452
devenv.exe	3277.051
DB Browser for SQLite.exe	2827.803
explorer.exe	2212.040

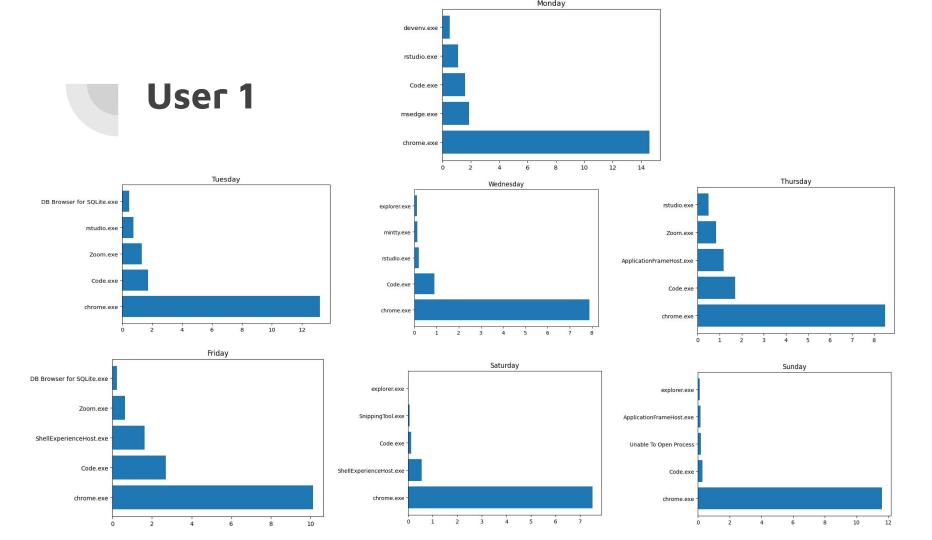


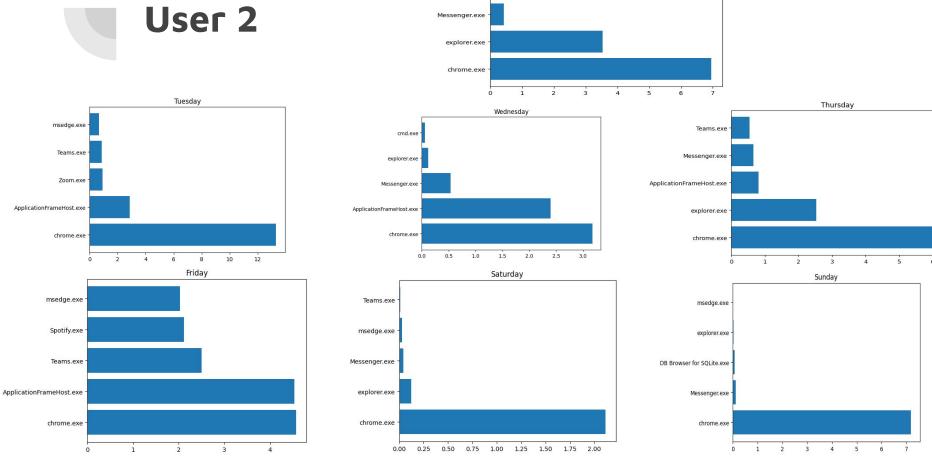
Data Preparation (User 2)



Data Visualization





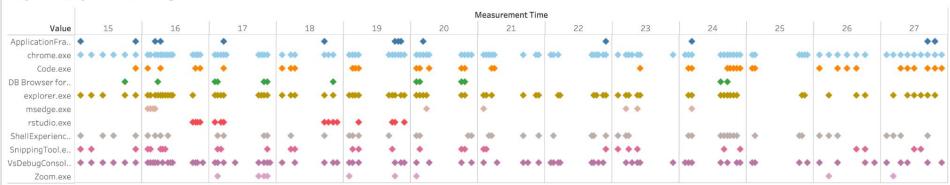


Teams.exe

Zoom.exe

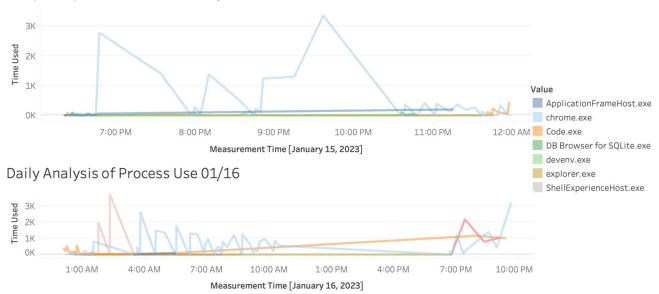
Monday





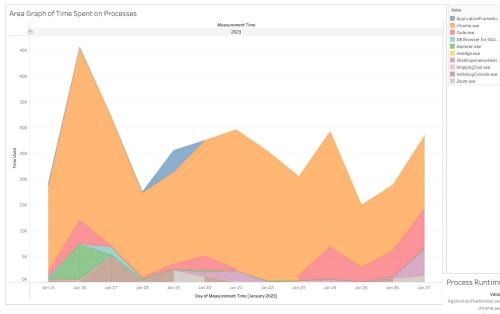
- A visual representation of the time spent on multiple processes from Jan 15 Jan 27, 2023
- A calendar-like representation / break down of top 10 most used processes by user 1

Daily Analysis of Process Use 01/15

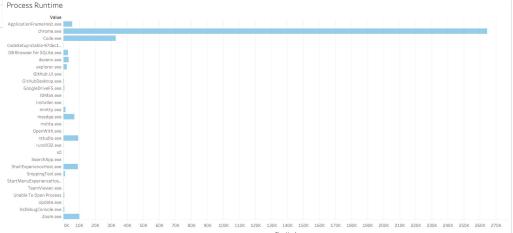


Daily Analysis of Process Use 01/18

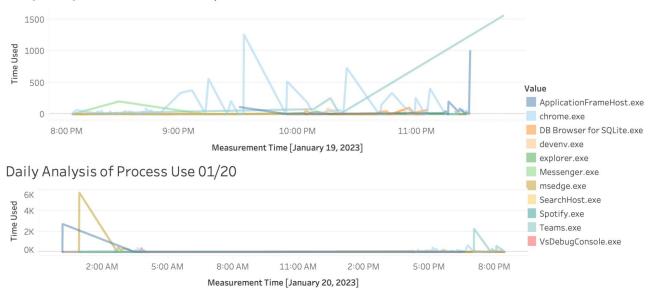




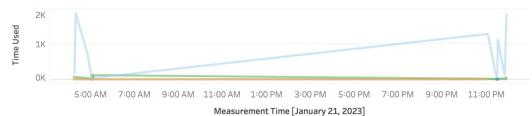
Average time spent actively using a particular application between 01/15/23 - 1/27/23



Daily Analysis of Process Use 01/19

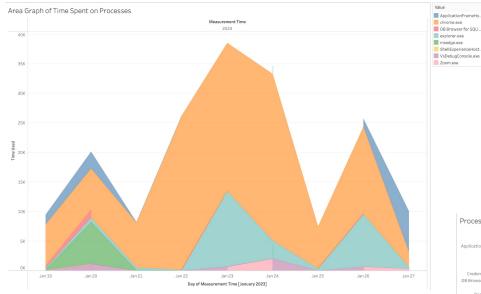


Daily Analysis of Process Use 01/21

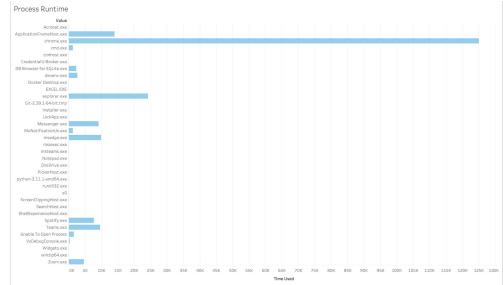




- A visual representation of the time spent on multiple processes from Jan 19 Jan 27, 2023
- A calendar-like representation / break down of top 10 most used processes by user 2



Average time spent actively using a particular application between 01/19/23 - 1/27/23



Problem statement

Main task:

Predict the total time an app/tab is used in the foreground

- **Subtask 1**: Predict #hrs an app is used during a day of a week
- Subtask 2: Predict #secs an app is used during an hour of a day
- ⇒ Time-series forecasting ⇒ RNN/LSTM

Feature Selection

- MEASUREMENT_TIME
 - ⇒ The *datetime* the app was used
 - ⇒ Use time in secs (of an hr) or hrs (of a day)

<u>Example</u>				
Арр 1	hr1 [12am, 1am)	hr2 [1am, 2am)	 hr23 [10pm, 11pm)	hr24 [11pm, 12am)
day_1	1800 sec	0 sec	900 sec	450 sec
day_2	900 sec	0 sec	3600 sec	3600 sec
day_n	1800 sec	450 sec	1000 sec	200 sec

<u>Example</u>						
Арр 1	Monday	Tuesday		Saturday	Sunday	
week_1	3 hrs	2 hrs		5 hrs	4 hrs	
week_2	4 hrs	4 hrs		2 hrs	3 hrs	
•••						
week_n	2 hrs	5 hrs		2 hrs	4 hrs	

- Split data: train/validation/test sets: 70/20/10 %
 - No shuffle before splitting data to ensure
 - Consecutive windows of samples
 - Realistic results from the validation/test sets
 - Evaluated on data collected after training models



- Normalize data:
 - Scale features before training
 - Sin/Cos transform the "time of day", "time of week" signals

 $x_{scaled} =$

- 12 am \rightarrow 12 am (again)
- Mon → Mon (again)
 - Periodically
- Min-Max scale the use time
 - Preserve the original distribution
 - Smaller STD, reduce the effect of outliers
- Z-score:

$$Z = \frac{x - \mu}{\sigma}$$

Do this separately for train/validation/test sets

- "Window" data:
 - Make a prediction (24*7*1) hrs into the future, given (24*7*7) hrs of history
 - ⇒ Our the window:

1176 hrs

168 hrs

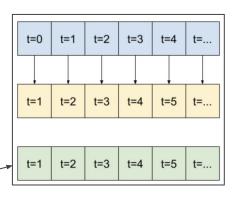
t=0 t=1 t=2 ... t=1175 t=1511 t=1512 ... t=1679

"WindowGenerator"
(if use Tensorflow)

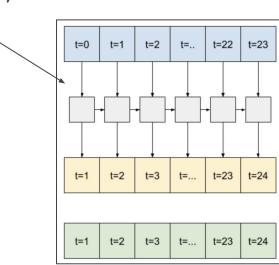
Validation Set Prediction Width

- Tune hyperparameters:
 - (Recurrent) Activation functions
 - tanh(x), relu, ...
 - Dropout; Regularization
 - Prevent overfitting
 - Bias for layers

- Compare the models:
 - Baseline
 - Dense layers
 - LSTM layers



- ~ linear model
- + dense layers



- Performance Metrics:
 - (Root) Mean Squared Errors
 - Mean Absolute Errors

$$RMS = \sqrt{rac{1}{n}\sum_{i}x_{i}^{2}}$$

$$ext{MAE} = rac{1}{n} \sum_{i=1}^n |x_i - x|$$

Thank you!

- References:
 - https://www.tensorflow.org/tutorials/structured_data/time_series
 - https://keras.io/api/layers/recurrent_layers/lstm/