# EP2420 - Project 1: Task 1, Task 2.1

### André Silva

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#### Task I

1. The dataset *VoD flashcrowd* provides us with 36633 samples of 1670 features, and 9 different types of targets. In Table 1, the statistics for the chosen features, as well as the target *DispFrames*, which represents the video frame rate, are displayed.

Feature name	Mean	Std Dev	Maximum	Minimum	25th percentile	90th percentile
3_cpu16sys	3.55	1.95	$1.50 \times 10^{1}$	0.00	0.00	0.00
23_RxPacktes	$3.15 \times 10^{2}$	$4.12 \times 10^{2}$	$2.37 \times 10^{3}$	0.00	$3.70 \times 10^{1}$	$4.70 \times 10^{1}$
$\beta_{-}frmpg.s$	$-1.15 \times 10^{1}$	$2.07 \times 10^{4}$	$1.08 \times 10^{5}$	$-7.49 \times 10^4$	$-5.31 \times 10^4$	$-4.55 \times 10^4$
$\theta_{-}cpu18_{-}.idle$	$9.97 \times 10^{1}$	$5.98 \times 10^{-1}$	$1.00 \times 10^{2}$	$9.50 \times 10^{1}$	$9.70 \times 10^{1}$	$9.80 \times 10^{1}$
2_cpu17usr	$6.03 \times 10^{1}$	$1.82 \times 10^{1}$	$9.20 \times 10^{1}$	0.00	8.00	$1.34 \times 10^{1}$
4_cpu15iowait	$4.65 \times 10^{-3}$	$7.21 \times 10^{-2}$	3.03	0.00	0.00	0.00
4_cpu5sys	$2.90 \times 10^{1}$	$1.08 \times 10^{1}$	$7.40 \times 10^{1}$	0.00	6.06	9.18
$\beta_{-}cswch.s$	$7.30 \times 10^{4}$	$1.79 \times 10^{4}$	$1.07 \times 10^{5}$	$7.82 \times 10^{3}$	$2.19 \times 10^{4}$	$2.65 \times 10^{4}$
$38\_TxBytes$	$2.79 \times 10^{6}$	$3.18 \times 10^{6}$	$1.96 \times 10^{7}$	0.00	$3.47 \times 10^{5}$	$4.63 \times 10^{5}$
$2\_dev8.0\_avgrq.sz$	$2.21 \times 10^{1}$	$5.78 \times 10^{1}$	$1.02 \times 10^{3}$	0.00	0.00	0.00
DispFrames	$2.20 \times 10^{1}$	4.32	$2.50 \times 10^{1}$	0.00	3.00	8.00

Table 1: Statistics for chosen features and target

The following list gives a short description of these features. This information was retrieved from linux manual pages for the command sar [1].

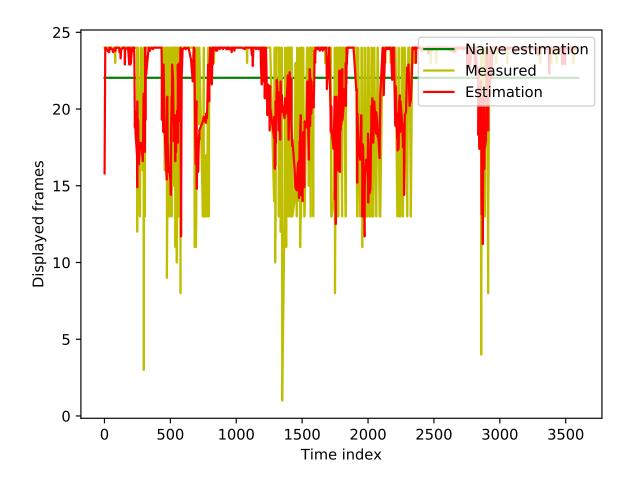
- 3\_cpu16.sys Percentage of CPU utilization that occurred while executing at the system level (kernel).
- 23\_RxPackets Total number of packets received per second.
- 3\_frmpg.s Number of memory pages freed by the system per second.
- 0-cpu18\_.idle Percentage of time that the CPU or CPUs were idle and the system did not have an outstanding disk I/O request.
- 2\_cpu17\_.usr Percentage of CPU utilization that occurred while executing at the user level (application).
- 4\_cpu15\_.iowait Percentage of time that the CPU was idle during which the system had an outstanding disk I/O request.
- 4-cpu5..sys Percentage of CPU utilization that occurred while executing at the system level (kernel).
- 3\_cswch.s Total number of context switches per second.
- $\bullet~38\_TxBytes$  Total number of bytes transmitted per second.
- 2\_dev8.0\_avgrq.sz The average size (in sectors) of the requests that were issued to the device.

### Task II - 2.1

4. Table 2 provides the calculated *Normalized Mean Absolute Error* (NMAE) for each of the regressors. The parameters utilized for the random forest regression and the neural network regression are the default parameters of, respectively, RandomForestRegressor [2] and MLPRegressor [3], unless specified in the table.

Regressor				
LinearRegression				
RandomForestRegressor(n_estimators=10)				
MLPRegressor(max_iter=1000, activation='logistic', hidden_layers=(10,10))	0.144			

Table 2: Normalized Mean Absolute Error for each regressor tested



 $\label{thm:cond} \mbox{Figure 1: Time series plot for } \textit{VoD flashcrowd} \ \mbox{using RnadomForestRegressor}$ 

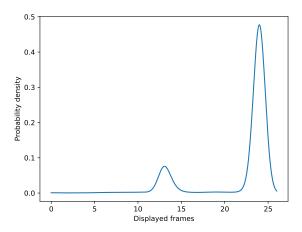


Figure 2: Density plot of the target values in the test set

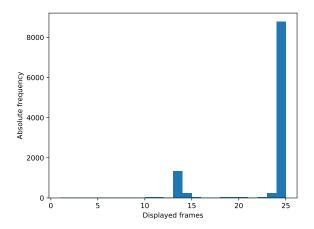


Figure 3: Histogram of the target values in the test set

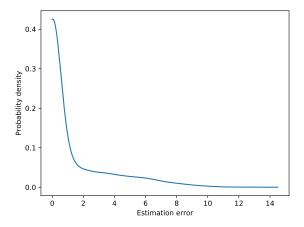


Figure 4: Density plot of the estimation errors  $y_i - \hat{y}_i$  in the test set

- 9. Of the three regression techniques utilized, linear regression was the least expensive in computational power, but also the one which provided the worst accuracy.
  - Neural network regression was more expensive when compared to random forest regression, but provided worse accuracy.

Random forest regression was better both in terms of accuracy and cost-performance trade-off when compared to the other two models.

## References

- [1] sar command manual page. https://linux.die.net/man/1/sar. accessed: 2020-10-31.
- [2] sklearn.ensemble.randomforestregressor. https://scikit-learn.org/stable/modules/generated/sklearn.ensemble.randomforestregressor.html#sklearn.ensemble.randomforestregressor.accessed: 2020-10-31.
- [3] sklearn.neural\_network.MLPRegressor. https://scikit-learn.org/stable/modules/generated/sklearn.neural\_network.MLPRegressor.html#sklearn.neural\_network.MLPRegressor.accessed: 2020-10-31.