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AI/ML Based Customer IO Workload Simulation

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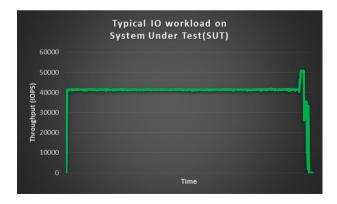
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- Objective
- Challenges
- ML Model
- Simulation
- Summary

Objective

- Test environment plays an important role in effective testing, thus reducing Customer Found Issues (CFIs)
- Simulation of customer like environment, especially the IO workload pattern for non-functional requirements, is an important aspect to consider.
- Typical IO Workload in test environment vs. customer environment

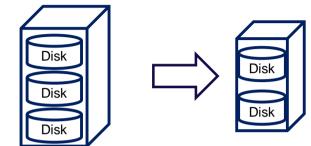




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Challenges

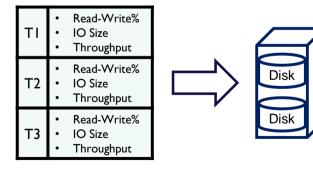




Configuration Incompatibility : IO Performance

depends on multiple factors – Storage Array configuration is one of them. Source array configuration most likely not the same as array under test.

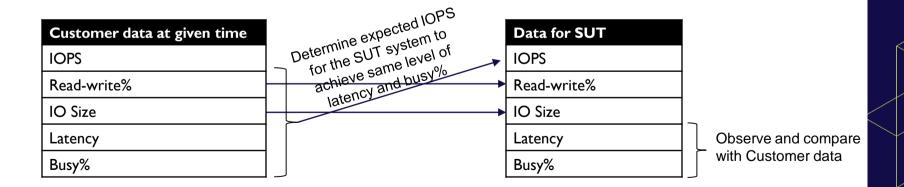
IO Workload Replay : Customer storage system performance varies along with time and tool must be capable of replaying the workload along with time.



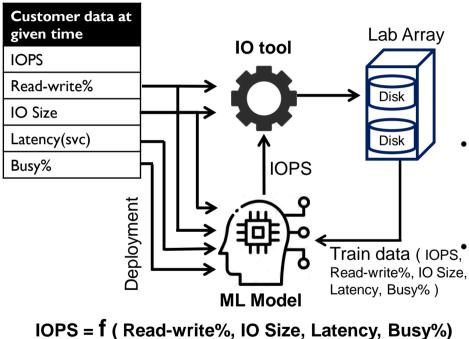
Workload Completeness : There are many factors than just one or more parameters of the workload characteristics does not make simulation complete.

Approach

- Get System Under Test(SUT) storage array in similar busy state as that of customer array for each data sample.
- Instead of using IOPS directly from customer performance data, determine expected IOPS for the SUT system for given Read-Write%, IO Size, Latency and Busy% from Customer data.



Machine Learning (ML) model



- Step1: To match load similar to source array on a different configuration, first ran different workloads on lab array to build ML model using standard python modules.
- **Step2**:Once the model is built, used customer performance data to determine required IOPS for SUT system.

Step3: Generated workload with all identified parameters on to SUT - Read-write%, IO Size, along with ML estimated IOPS.

Step4: Repeat same with each data sample for all time slots.

Accuracy of the ML Model

In [147]:	y_pred = s2286_model.	predict(X_test)	<pre>In [157]: from sklearn.metrics import mean_squared_error</pre>
In [148]:	<pre>y_df = pd.DataFrame(</pre>	{ "actual": y_test, "predicted": y_pred,	<pre>In [158]: mse_s2286_model = mean_squared_error(y_df.actual,</pre>
		<pre>"residual": y_test - y_pred })</pre>	In [159]: import numpy as np
In [149]:	y_df		<pre>In [160]: np.sqrt(mse_s2286_model)</pre>
Out[149]:	actual predicted	residual	Out[160]: 75.39296915165409
	1 4947.1 4798.818983	148.281017	In [161]: from sklearn.metrics import r2_score
	30 3083.2 3057.271086	25.928914	
	31 3019.8 3116.371941	-96.571941	In [162]: r2_score(y_df.actual, y df.predicted)
	36 2106.7 2133.335032	-26.635032	Out[162]: 0.9934323420220152
	23 2947.2 2953.389564	-6.189564	K
	37 2059.2 1972.005738	87.194262	<pre>In [150]: customer_df = pd.read_csv("./Input_data/customer.csv")</pre>
	8 3971.6 3921.641787	49.958213	
	13 2281.0 2332.387662	-51.387662	 This set of data generated on SUT system only used fo testing the model. Model was not built with these data
	30 3083.2 3057.271086 31 3019.8 3116.371941 36 2106.7 2133.335032 23 2947.2 2953.389564 37 2059.2 1972.005738 8 3971.6 3921.641787 13 2281.0 2332.387662	25.928914 -96.571941 -26.635032 -6.189564 87.194262 49.958213	<pre>In [162]: r2_score(y_df.actual, y_df.predicted) Out[162]: 0.9934323420220152 In [150]: customer_df = pd.read_csv("./Input_data/customer.csv"</pre>

In [150]: customer df = pd.read csv("./Input data/customer.csv")

Mean error between actual test data and predicted data is just 75 IOPS.

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R2-score value is between 0.0 and 1.0. Where 0.0 is being highly incorrect model and 1.0 indicates perfect model – In this case, it shows .99 with test data set

Prediction Vs. Simulation

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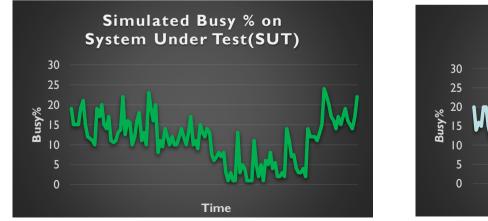
In [155]:	custom	er_df['t	gt_iops'													
In [156]:	custom	Customer df Actual IOPS on Estimated IOPS														
Out[156]:	Customer array on Lab array											Lab array				
000[100].		Wr_iops	Tot_iops	Rd_kbps	Wr_kbps	Tot_kbps	Rd_svc	Wr_svc	Tot_svc	Rd_iosz	Wr_iosz	Tot_iosz	QLen	AvgBusy%	Rd_Wr_ratio	tgt_iops
	1449.5	619.8	2069.3	11873.9	5077.7	16951.6	0.433	0.333	0.403	8.2	8.2	8.2	1	20.8	2.338658	5061.586494
Ζ	588.5	1232.6	1821.1	11530.3	37052.3	48582.6	0.623	0.390	0.465	19.6	30.1	26.7	2	21.2	0.477446	4787.793345
Í I	344.0	1379.0	1723.0	11272.6	45185.8	56458.4	0.878	0.396	0.492	32.8	32.8	32.8	2	21.2	0.249456	4734.335843
	339.6	1353.9	1693.5	11128.4	44365.9	55492.3	0.929	0.397	0.504	32.8	32.8	32.8	0	21.3	0.250831	4698.181568
4 1	340.5	1362.2	1702.7	11157.1	44636.9	55794.0	0.934	0.396	0.504	32.8	32.8	32.8	1	21.4	0.249963	4680.126253
	304.3	1213.1	1517.4	9969.9	39751.6	49721.5	0.967	0.401	0.515	32.8	32.8	32.8	1	19.5	0.250845	5046.749119
	329.2	1319.3	1648.6	10788.7	43231.8	54020.5	1.012	0.399	0.521	32.8	32.8	32.8	2	21.5	0.249526	4630.174092
	274.3	1100.2	1374.4	13439.2	54057.3	67496.6	1.371	0.457	0.639	49.0	49.1	49.1	1	22.0	0.249318	4306.553536
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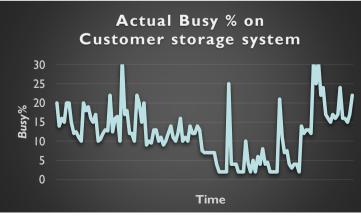
Time Wr Wr Wr Tot QLen AvgBusy% Secs Rd Tot Rd Tot Rd Wr Tot Rd 2020-09-02 09:35:00 PDT 1591115700 986.6 3949.9 4936.5 32360.4 129556.7 161917.2 0.631 0.374 0.405 32.8 32.8 32.8 19.7 0 2020-09-02 09:40:00 PDT 1591116000 928.5 3819.8 4748.3 30454.8 125289.4 155744.2 0.866 0.372 0.511 32.8 32.8 32.8 20.2 1 2020-09-02 09:45:00 PDT 1591116300 898.3 3609.9 4508.2 29464.2 118404.7 147868.9 0.582 0.364 0.416 32.8 32.8 32.8 20.8 0 2020-09-02 09:50:00 PDT 1591116600 951.6 3994.1 4945.7 31212.4 131006.4 162218.9 0.699 0.562 0.419 32.8 32.8 32.8 2 20.1 20.5 4 932.7 3718.4 4651.1 30592.5 121963.5 152556.1 0.668 0.374 0.463 32.8 32.8 32.8 0.8

root@mnodee117c885:~#

Busy% of Lab vs Customer Array

- Existing internal IO tool has been enhanced to supply IOPS as an input along with other parameters and iterate every data samples to replay the workload.
- Following graphs shows that simulated load on System under test is very close to the one in Customer storage system:
- Able to reach close to Busy% and IO processing(Latency) condition on Lab array with same as customer workload characteristics





Thank You for Attending

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