

# SAS OPTIMIZATION CHALLENGE

Report

**SAS Poets**

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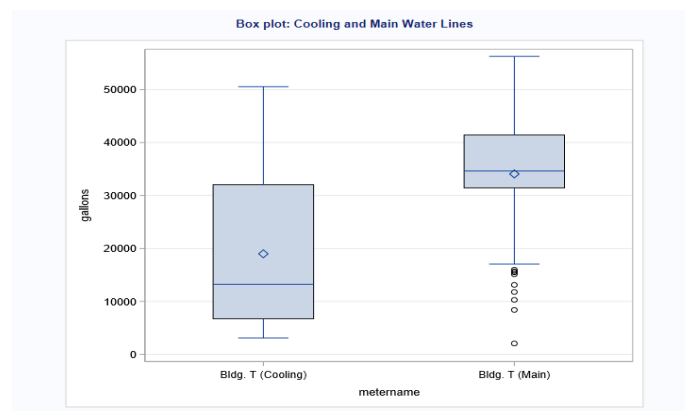
## Problem Statement

Building T in XYZ Corporation meets its water requirements from two sources, the Water Co. and its own water storage tank. Based on a year and a half of historical weekly gallon usage of water, we need to predict their water requirements for the next 4 weeks. To meet the predicted requirements, XYZ has two contract options at hand, each involving a certain cost. We need to optimize the cost function to help XYZ decide which contract to opt for. There are multiple constraints like water in the storage tank should not drop below 30000 gallons at any point, at least 25% of the water requirements should be met from the water storage tank, etc., which needs to be met while optimizing the cost.

## Forecasting the requirements for next 4 weeks

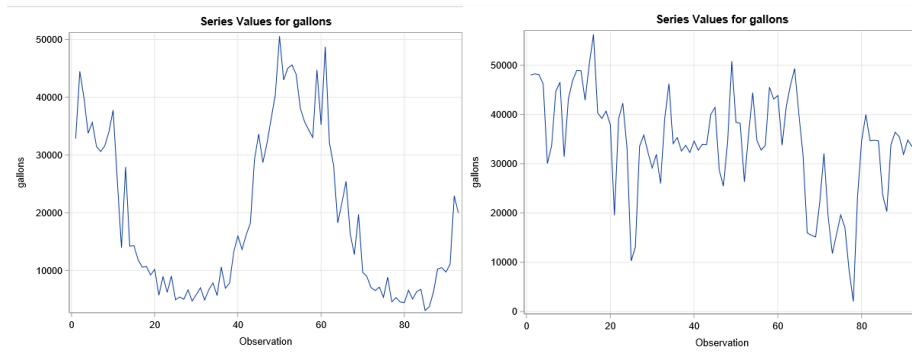
## Exploratory Data Analysis

We observed that the distributions for cooling and main types are significantly different through our initial data analysis. Hence they needed to be predicted through different models and added up to calculate the final demand.



Here we can observe that main has more water requirements than cooling and less variance than cooling.

**Time series graph for Cooling and Main:**



We can observe that cooling peaks during summer whereas main dips during summer.

ARIMA model has been used to forecast the water requirements for the next 4 weeks.

#### Forecasting for Main

Name of Variable = gallons	
Mean of Working Series	34075.47
Standard Deviation	10681.6
Number of Observations	93

Autocorrelation Check for White Noise								
To Lag	Chi-Square	DF	Pr > ChiSq	Autocorrelations				
6	97.58	6	<.0001	0.672	0.356	0.286	0.381	0.357
12	105.29	12	<.0001	0.177	0.140	0.122	0.062	0.064
18	112.14	18	<.0001	-0.133	-0.141	-0.058	-0.022	-0.055

Maximum Likelihood Estimation					
Parameter	Estimate	Standard Error	t Value	Approx Pr >  t	Lag
MU	34420.5	2782.4	12.37	<.0001	0
AR1,1	0.88136	0.07425	9.18	<.0001	1
AR2,1	0.26506	0.13661	1.90	0.0576	52

Constant Estimate	8080.67
Variance Estimate	59232810
Std Error Estimate	7696.272
AIC	1935.705
SBC	1943.302
Number of Residuals	93

Correlations of Parameter Estimates				
Parameter	MU	AR1,1	AR2,1	
MU	1.000	-0.009	-0.109	
AR1,1	-0.009	1.000	-0.015	
AR2,1	-0.109	-0.015	1.000	

Autocorrelation Check of Residuals								
To Lag	Chi-Square	DF	Pr > ChiSq	Autocorrelations				
6	13.47	4	0.0092	0.103	-0.227	-0.129	0.208	0.121
12	15.89	10	0.1028	0.027	0.018	0.048	-0.070	0.110
18	20.79	16	0.1887	-0.130	-0.124	0.044	0.080	0.032
24	33.10	22	0.0605	-0.164	0.056	0.138	0.054	-0.103

#### Forecasting for Cooling

Name of Variable = gallons	
Mean of Working Series	18991.63
Standard Deviation	13805.28
Number of Observations	93

Autocorrelation Check for White Noise								
To Lag	Chi-Square	DF	Pr > ChiSq	Autocorrelations				
6	364.24	6	<.0001	0.915	0.869	0.813	0.759	0.699
12	444.10	12	<.0001	0.549	0.471	0.389	0.285	0.177
18	486.81	18	<.0001	-0.018	-0.098	-0.175	-0.249	-0.322

Maximum Likelihood Estimation					
Parameter	Estimate	Standard Error	t Value	Approx Pr >  t	Lag
MU	20140.1	5646.7	3.57	0.0004	0
AR1,1	0.89839	0.04470	20.10	<.0001	1
AR2,1	0.24545	0.15695	1.53	0.1249	52

Constant Estimate	1544.143
Variance Estimate	28498110
Std Error Estimate	5338.362
AIC	1868.129
SBC	1875.727
Number of Residuals	93

Correlations of Parameter Estimates				
Parameter	MU	AR1,1	AR2,1	
MU	1.000	-0.037	0.032	
AR1,1	-0.037	1.000	-0.172	
AR2,1	0.032	-0.172	1.000	

Autocorrelation Check of Residuals								
To Lag	Chi-Square	DF	Pr > ChiSq	Autocorrelations				
6	11.90	4	0.0181	-0.281	0.123	0.084	-0.025	0.176
12	17.40	10	0.0660	0.029	0.173	-0.002	-0.041	0.109
18	19.45	16	0.2458	-0.042	0.004	-0.040	0.008	0.013
24	25.70	22	0.2648	-0.053	-0.094	-0.016	-0.087	-0.007

Model for variable gallons				
Estimated Mean		20140.05		

Autoregressive Factors	
Factor 1:	1 - 0.88839 B <sup>1</sup> (1)
Factor 2:	1 - 0.24545 B <sup>1</sup> (52)

Forecasts for variable gallons				
Obs	Forecast	Std Error	95% Confidence Limits	
94	20442.2901	5338.3621	9979.2927	30905.2875
95	20804.8350	7178.2836	6739.5777	34870.0923
96	23406.2825	8370.3758	7000.6274	39811.8978
97	24387.4799	9222.0597	6282.5750	42442.3848

Outlier Detection Summary	
Maximum number searched	2
Number found	2
Significance used	0.05

Outlier Details				
Obs	Type	Estimate	Chi-Square	Approx Prob>ChiSq
61	Additive	15282.0	28.87	<.0001
13	Additive	13804.0	27.29	<.0001

Model for variable gallons				
Estimated Mean		34420.53		

Autoregressive Factors	
Factor 1:	1 - 0.88136 B <sup>1</sup> (1)
Factor 2:	1 - 0.28508 B <sup>1</sup> (52)

Forecasts for variable gallons				
Obs	Forecast	Std Error	95% Confidence Limits	
94	34404.3848	7898.2725	19319.9480	49488.7816
95	34353.4828	9312.9683	16100.3803	52606.5453
96	35958.4281	9974.8532	16408.0750	55508.7812
97	36308.9950	10267.642	16184.7570	56433.1730

Outlier Detection Summary	
Maximum number searched	2
Number found	2
Significance used	0.05

Outlier Details				
Obs	Type	Estimate	Chi-Square	Approx Prob>ChiSq
21	Additive	-16844.4	7.24	0.0071
67	Shift	-10390.0	7.78	0.0053

**Final Prediction for the following 4 weeks:**

[1]	gallons
1	54847
2	55158
3	59365
4	60676

## Business Insights

How many gallons will XYZ buy from The Water Co. each week?

Week	WaterCo Quantity used
1	41,135
2	35,000
3	35,000
4	35,000
Total	146,135

How many gallons will XYZ use from their Water Storage Tank each week?

Week	Water Tank Quantity used
1	13,712
2	20,158
3	24,365
4	25,676
<b>Total</b>	<b>83,911</b>

What is XYZ's projected total water cost at the end of the next four weeks?

Week	Water Tank Cost
1	2,468.1
2	3,628.5
3	2,436.5
4	2,567.6
<b>Total</b>	<b>11,100.7</b>

What is XYZ's projected ending Water Storage Tank inventory at the end of each week?

Week	Inventory = Prev Weeks inventory + precipitation - Amount from tank used by building T	Precipitation	Amount from tank used by building T
1	60,788	12,000	13,712
2	58,630	18,000	20,158
3	56,265	22,000	24,365
4	52,589	22,000	25,676
<b>Total</b>	<b>228,272</b>	<b>74,000</b>	<b>83,911</b>

How much money will XYZ save by choosing the recommended contract over the alternative contract?

**Recommended Contract: Contract B**

Week	Total Cost
1	7,404.3
2	7,828.5
3	6,636.5
4	6,767.6
<b>Total</b>	<b>28,637</b>

#### Alternative Contract: Contract A:

Week	Total Cost Used
1	8,638.3
2	8,687.4
3	7,186.5
4	7,317.6
Total	31,829.8

We will end up saving \$3,129 which is 10% lesser than what contract A would end up costing.

How many more/less gallons will be in the Water Storage Tank at the end of the four-week period compared to if the alternative contract was chosen?

Week	Inventory = Prev Weeks inventory + precipitation - Amount from tank used by building T	Precipitation	Amount from tank used by building T
1	60,788	12,000	1,712
2	64,999	18,000	13,790
3	52,634	22,000	34,365
4	38,958	22,000	35,676

## Recommendations

We recommend XYZ Co. to go ahead with Contract B as Contract 2 is 10 % cheaper than Contract A.

We can use our forecasting model to predict water needs for longer duration so that we can negotiate better with Water Co. and we can initiate a bidding process to invite more bids which would be evaluated using our optimization model. That way we can save more cost for XYZ Co.