



Measuring energy demand and efficiency at WWTPs: an econometric approach

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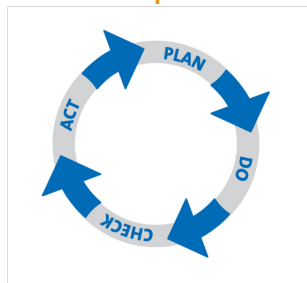
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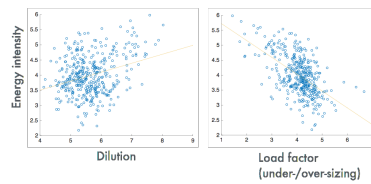
INTRODUCTION

Increasing energy efficiency is a cyclic process based on continually performance measurement (benchmarking). However, measuring energy efficiency at WWTPs is still a challenge.

"If you can't measure it, you can't improve it"



Energy use is affected by exogenous variability



Energy efficiency is typically approximated by energy intensity, which is defined as the amount of energy use per unit of activity (e.g. volume of treated wastewater). Changes in energy intensity are just approximate indicators for changes in energy efficiency since they are affected by external factors such as the influent characteristics, climate, scale effect of plant size and other construction parameters.

Furthermore, considering that WWTPs perform different functions, i.e. removing of COD, removing of N and/or P, resource recovery, producing an effluent free of pathogens, general energy intensity indicators (i.e. kWh/m³ or kWh/PE) have limited value, as they do not provide enough information of the WWTPs operation.

WWTPs can have different functions

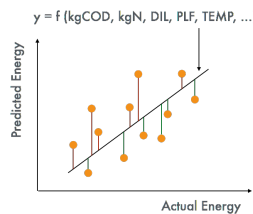
$$\text{Energy efficiency} = \frac{\text{Input}}{\text{Output}}$$

Input: kWh
Output: m³, kgCOD, kgN, kgP, pathogen

METHODOLOGY

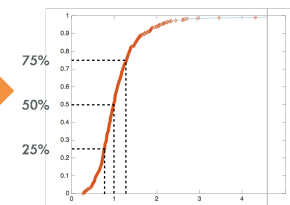
In this study, a linear regression approach is used to control for aspects that systematically influence the energy use at WWTPs. After estimating the log-log model, the β coefficients can be used to determine the impact of independent variables (X) on dependent variable (Y). Moreover a benchmark table can be build from the ratios between predicted and actual energy consumption.

Ordinary least square (OLS) predicts the expected energy consumption



$$\text{EE Ratio} = \frac{\text{Predicted}}{\text{Expected}}$$

Benchmark table from cumulative distribution of EE ratios



RESULTS AND CONCLUSIONS

OLS for understanding

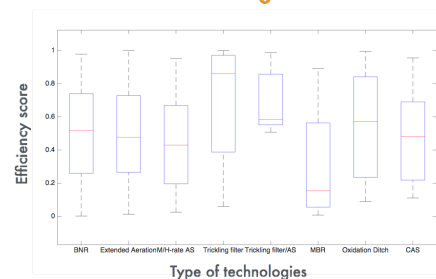
Estimation results

Parameter	M1 [Y=kWh/day]	M2 [Y=kWh/m ³]	M3 [Y=kWh/PE]
Intercept	-0.403***	-0.059	-0.835***
SacTreat (Trickling filter)			-0.122
BNR	0.430***	-0.043	0.893***
Extended Aeration	0.420***	-0.046	0.871***
M/H rate AS	0.317***	-0.078	0.659***
Trickling filter-AS	0.330***	-0.125	0.684***
MBR	0.548***	-0.112	1.138***
Oxidation Ditch	0.316***	-0.09	0.655***
CAS	0.370***	-0.119	0.768***
TerTreat (NO)			-0.247
YES	0.132**	-0.051	0.275**
InTemp	0.084***	-0.021	0.174***
InSIZE	0.912***	-0.018	-0.348***
InPLF	-0.114***	-0.015	-0.237***
InDIL	0.038**	-0.018	-0.599***
InNout	0.095***	-0.015	-0.197***
Root Mean Squared Error	0.273	0.565	0.711
Adjusted R-Squared	0.926	0.681	0.496

*** Significant at 1% level; ** Significant at 5% level; * Significant at 10% level.

OLS for benchmarking

Are there differences across technologies?



- Following the previously described procedures three regression models were tested in order to describe the relationship between energy consumption and operational parameters and the results are reported in the following table.

- OLS is a robust technique that can take into account the large heterogeneity which WWTPs are characterized.
- Thanks to the flexibility on the inclusion of different type of variables this technique can be used to test for hypothesis.