



# Design, Benchmarking and Graphical Lasso based Explainability Analysis of an Energy Game-Theoretic Framework

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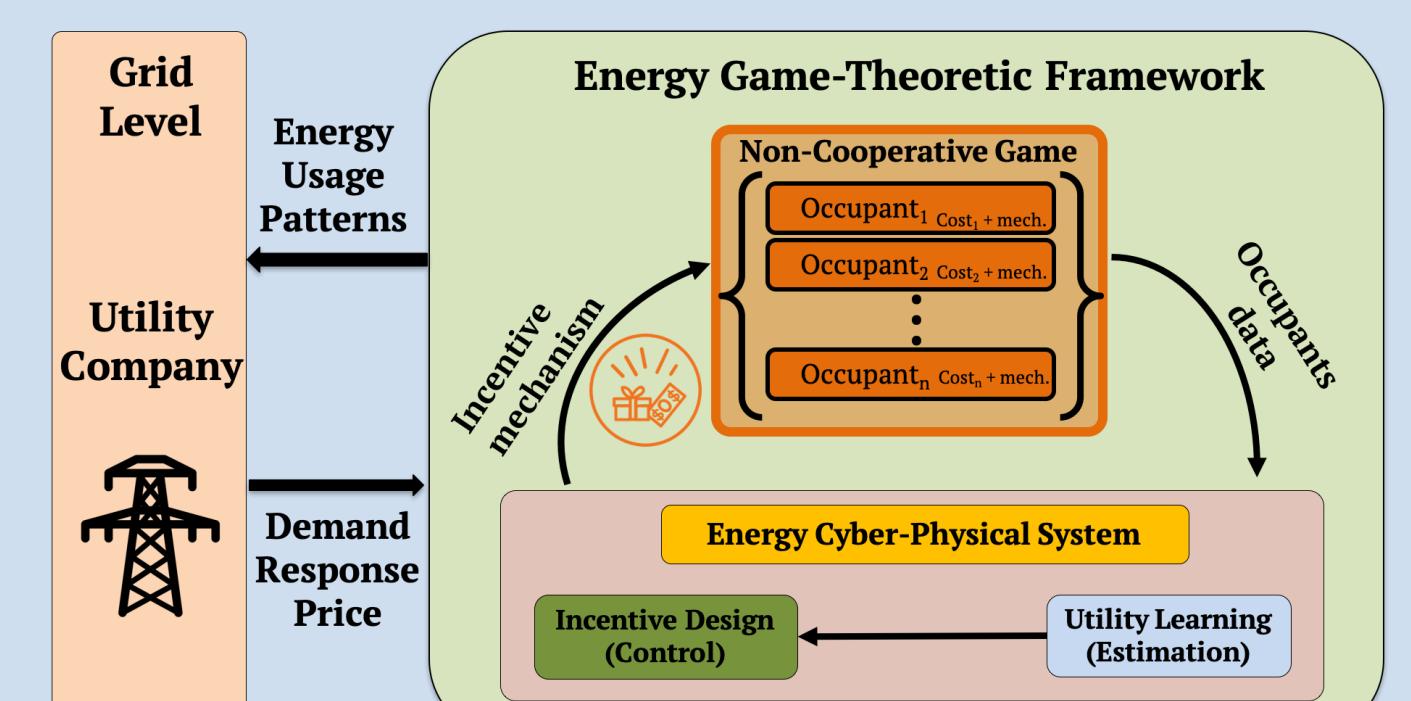


## Introduction

Buildings, both residential and commercial, account for more than  $\sim 50\%$  of global electricity consumption and  $\sim 40\%$  of worldwide  $CO_2$  emissions!

- Attempts to improve energy efficiency in buildings include implementing control and automation approaches alongside techniques like incentive design and price adjustment to more effectively regulate the energy usage.
- But, occupants typically lack the independent motivation necessary to optimize their energy usage.
- Energy Game-Theoretic Framework:** Incentivize occupants to modify their behavior in a competitive game setting so that the over-all energy consumption in the building is reduced.
- The framework can also be integrated with the power grid to have dynamic protocols for demand response.

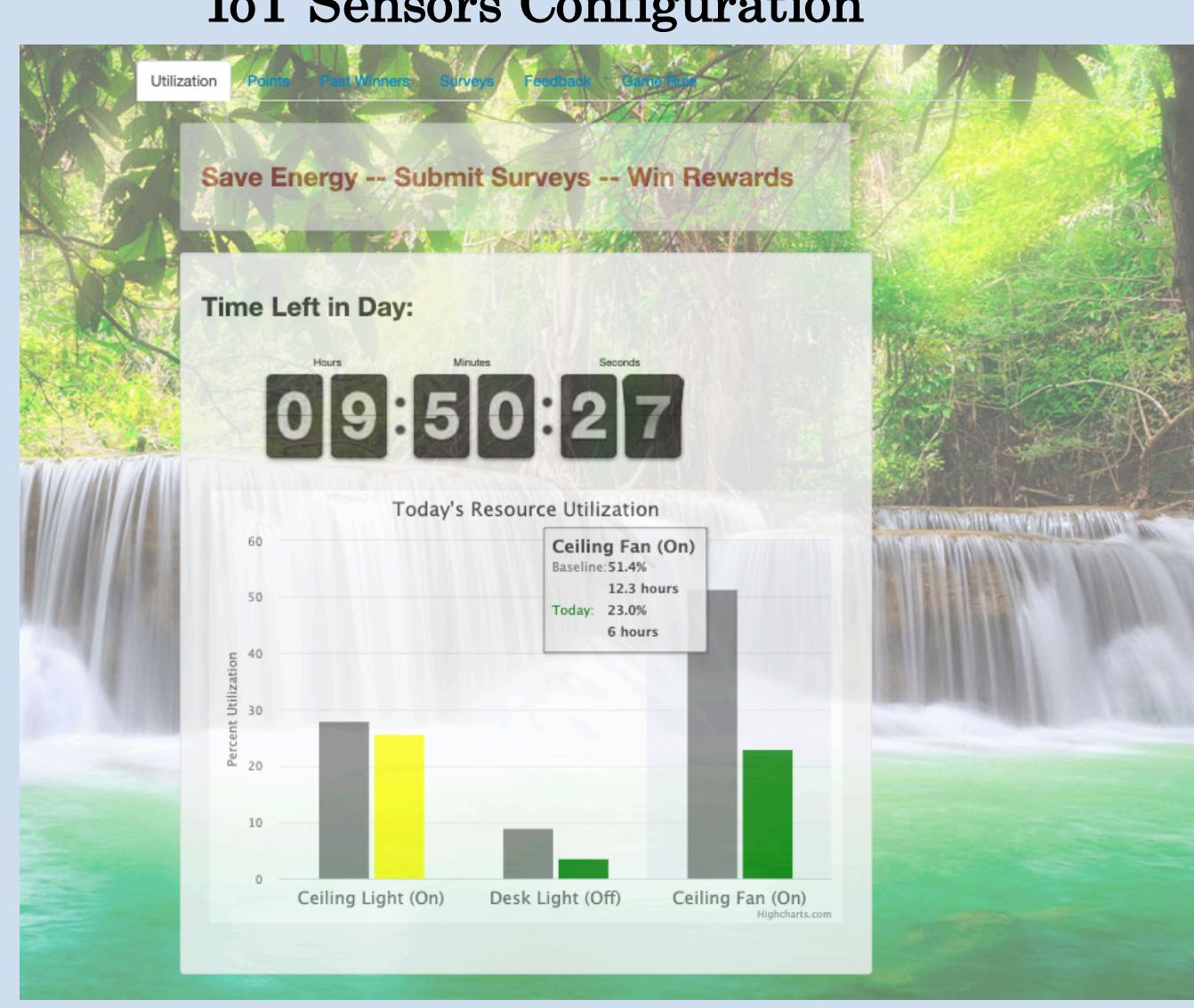
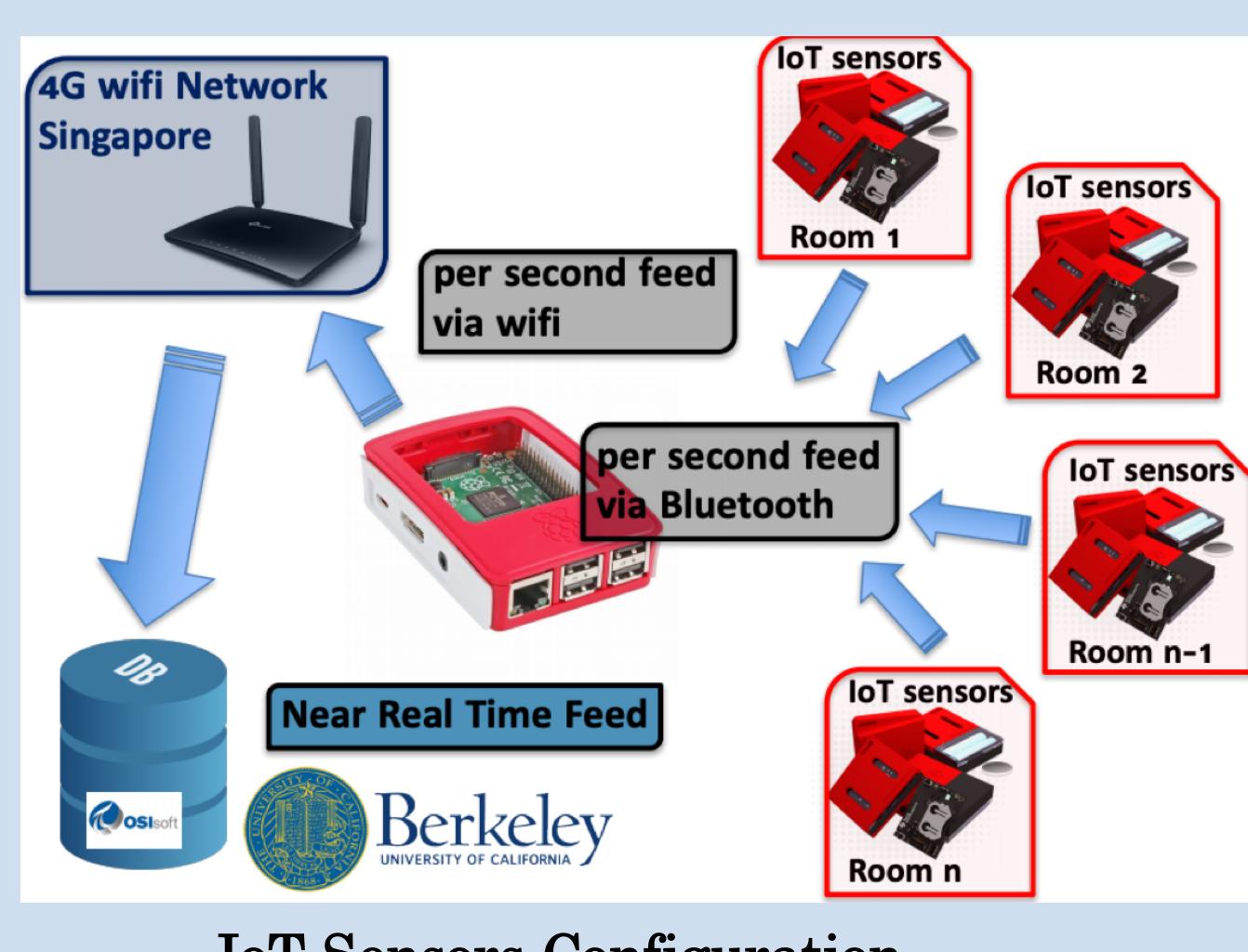
We present a benchmarked open-sourced dataset from an energy social game experiment at Nanyang Technological University (NTU), Singapore.



Interplay between electric grid and proposed framework

## Social Game Experiment

- Experimental environment: Residential housing single room apartments at NTU
- Deployed IoT sensors for energy resource observation and employed an web-interface for interaction with players



- The front-end was a web portal to report the occupants about real-time status
- Occupants were observed before game for one month, which serves as our baseline
- We employed a lottery mechanism consisting of gift cards to incentivize occupants, where the probability of winning was proportional to the players points in the game,

$$\hat{p}_i^d(b_i, u_i) = s_i \frac{b_i - u_i^d}{b_i}$$

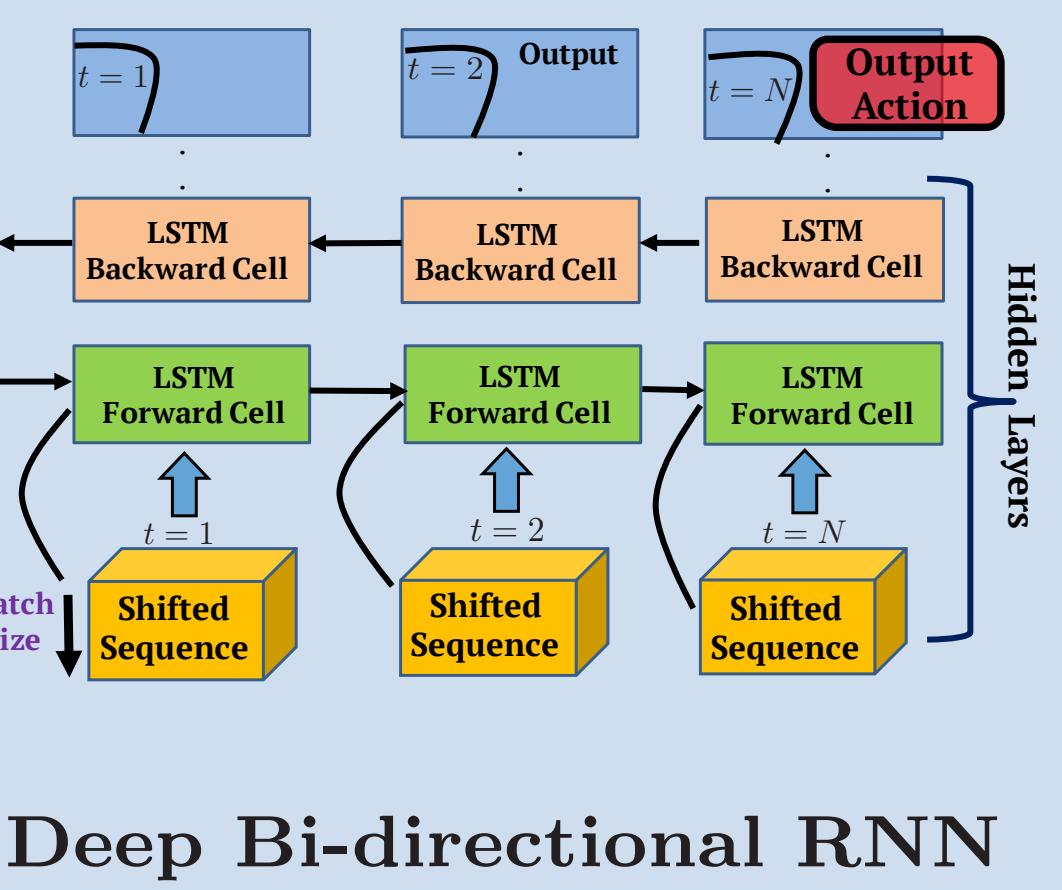
where  $\hat{p}_i^d$  is the points earned,  $u_i^d$  is the usage on day  $d$  for resource  $i$ ,  $b_i$  is the resource's baseline and  $s_i$  is a points booster for inflating the points as part of framing.

- We use sequential non-cooperative discrete game concept for the game design [1].

## Benchmarking of Energy Resource Usage Forecast

Since human interaction data in general is imbalanced, we use the Synthetic Minority Over-Sampling(SMOTE) technique for providing balanced data sets first. Reported are AUC scores for various models with step ahead (sensory data accounted) vs sensor free cases.

"Step-ahead"/"Sensor-free"	Ceiling Fan	C. Light	Desk Light
Logistic regression	0.83 / 0.65	0.78 / 0.61	0.78 / 0.68
Penalized $l_1$ Logistic regression	0.80 / 0.65	0.77 / 0.56	0.78 / 0.64
Bagged Logistic regression	0.84 / 0.66	0.80 / 0.59	0.79 / 0.68
LDA	0.81 / 0.65	0.78 / 0.58	0.74 / 0.68
K-NN	0.76 / 0.53	0.77 / 0.56	0.74 / 0.55
Support Vector Machine	0.82 / 0.65	0.78 / 0.60	0.76 / 0.68
Random Forest	0.91 / 0.60	0.78 / 0.59	0.98 / 0.63
Deep Neural Network	0.80 / 0.55	0.76 / 0.60	0.78 / 0.64
Deep Bi-directional RNN	<b>0.97 / 0.71</b>	<b>0.85 / 0.66</b>	<b>0.99 / 0.76</b>

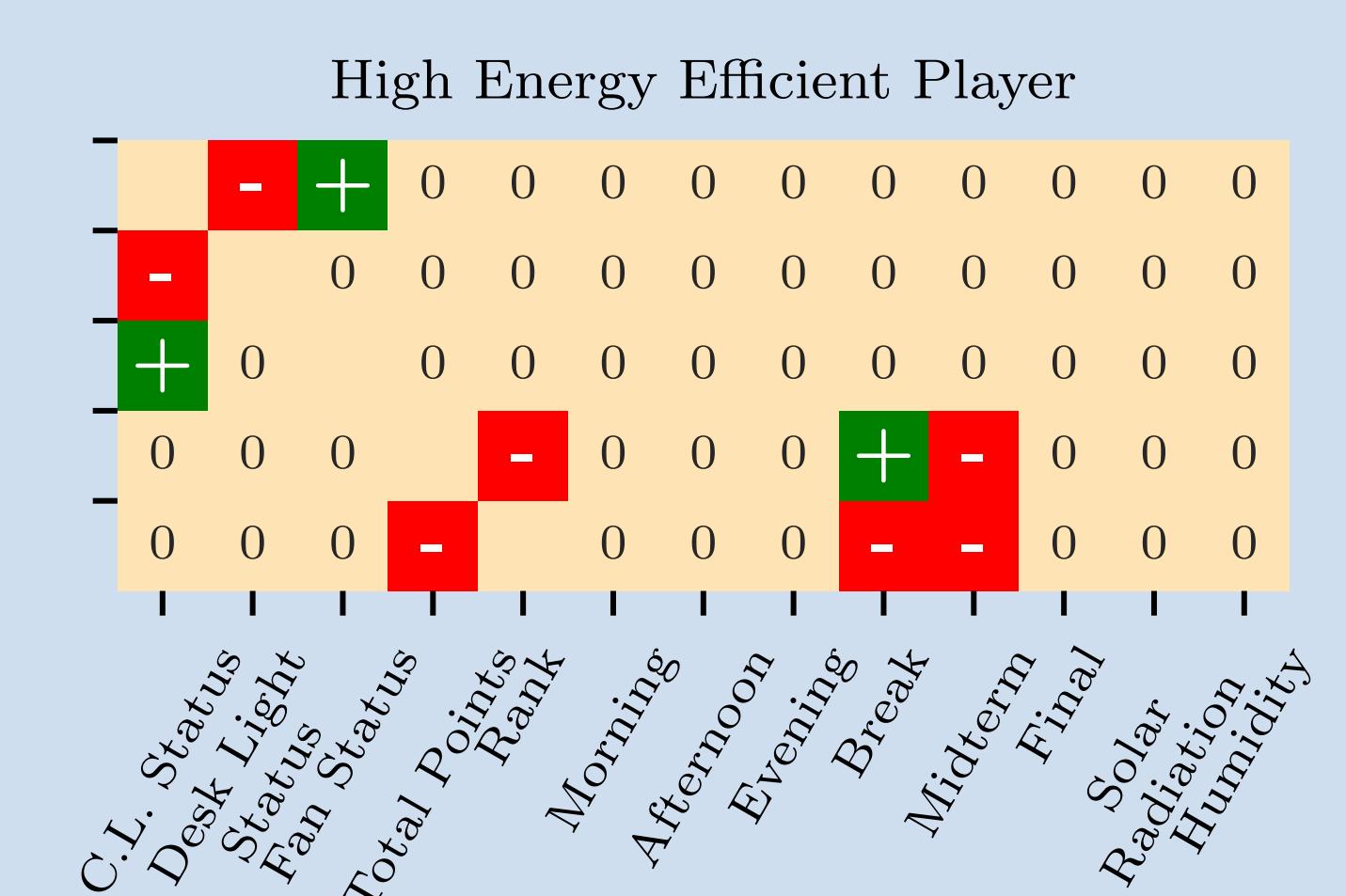
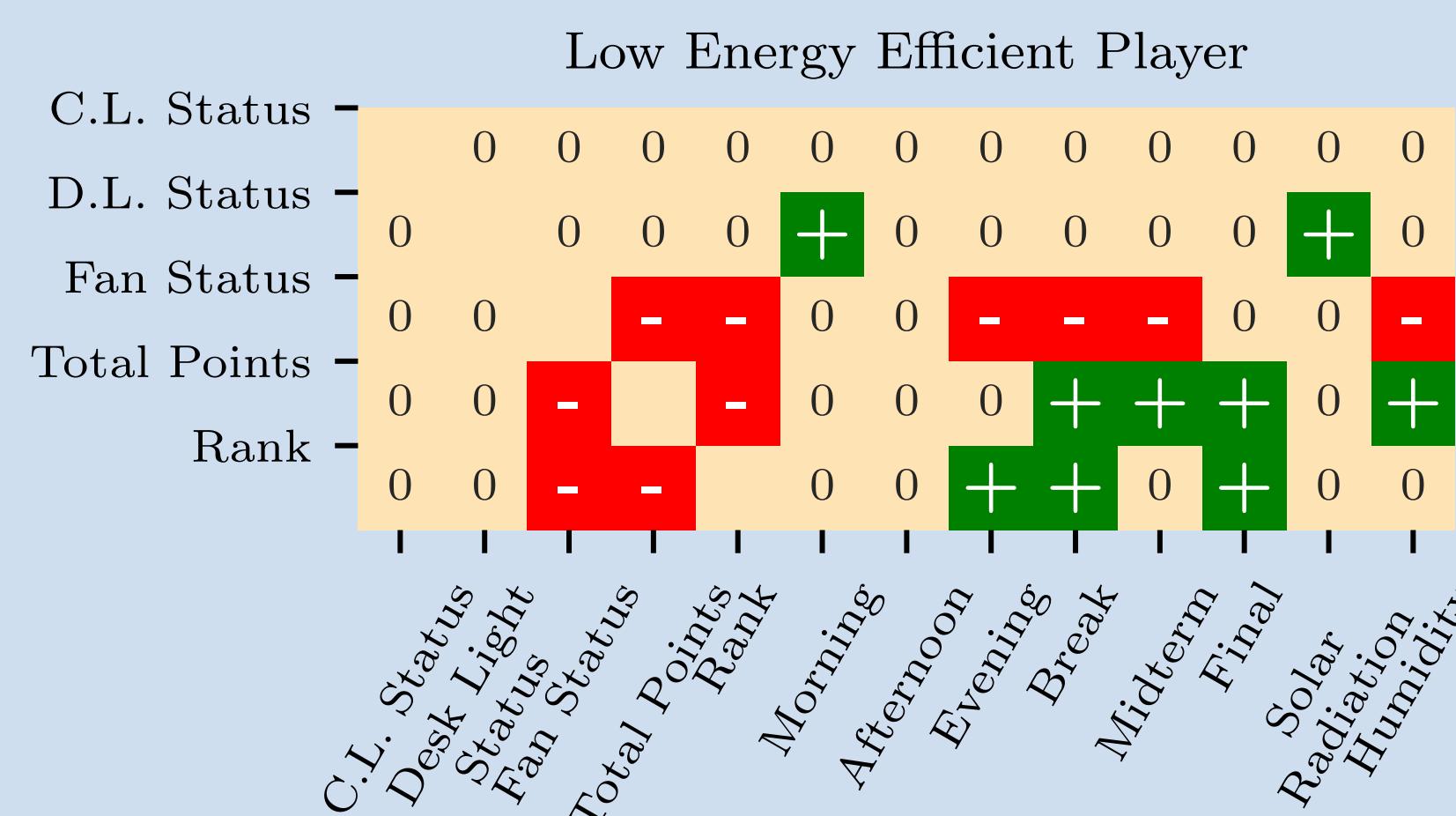


## Energy Savings achieved in the Social Game

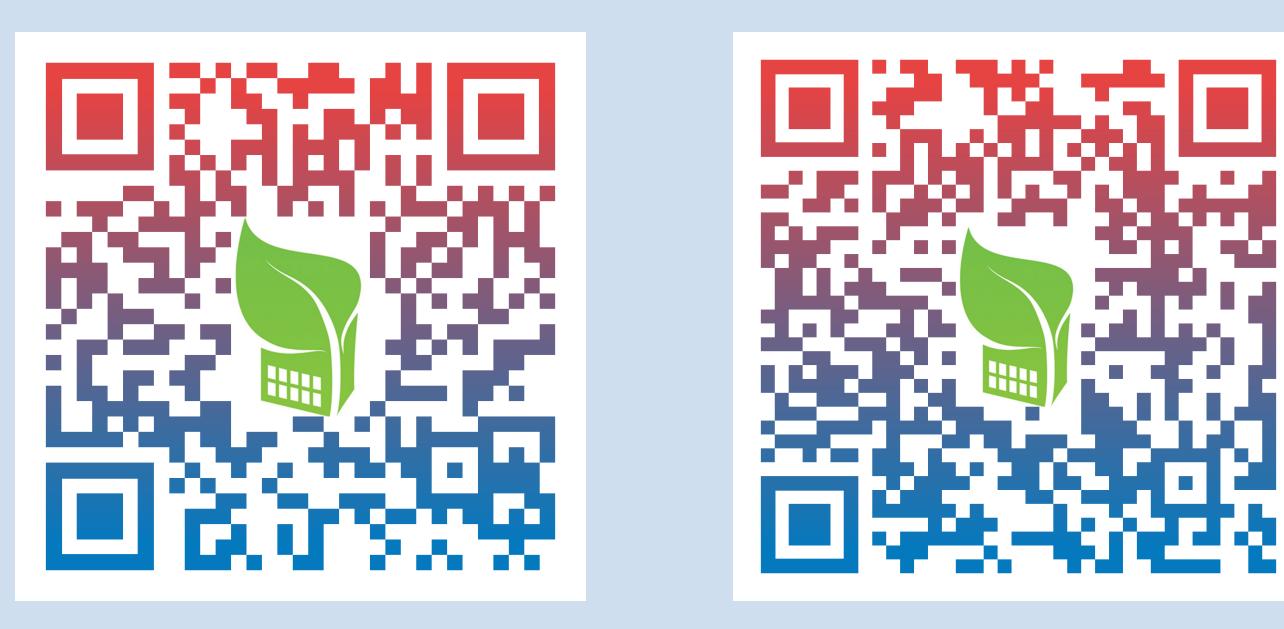
Device	Weekday				Weekend			
	Before	After	p-value	$\Delta$ %	Before	After	p-value	$\Delta$ %
Ceiling Light	417.5	393.9	0.02	5.6	412.3	257.5	0	37.6
Desk Light	402.2	157.5	0	60.8	517.6	123.3	0	76.2
Ceiling Fan	663.5	537.6	0	19.0	847.1	407.0	0	51.9

## Feature Correlation Learning using Graphical Lasso<sup>[1,2]</sup>

+ Positive Correlation    - Negative Correlation    0 No Correlation (correlation value not comparable with + or - correlation values)



## Open-Sourced Dataset



## References

- I. C. Konstantakopoulos et al. Design, Benchmarking and Explainability Analysis of a Game-Theoretic Framework towards Energy Efficiency in Smart Infrastructure. arXiv preprint arXiv:1910.07899, 2019
- H. P. Das et al. A Novel Graphical Lasso based approach towards Segmentation Analysis in Energy Game-Theoretic Frameworks. arXiv preprint arXiv:1910.02217, 2019

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