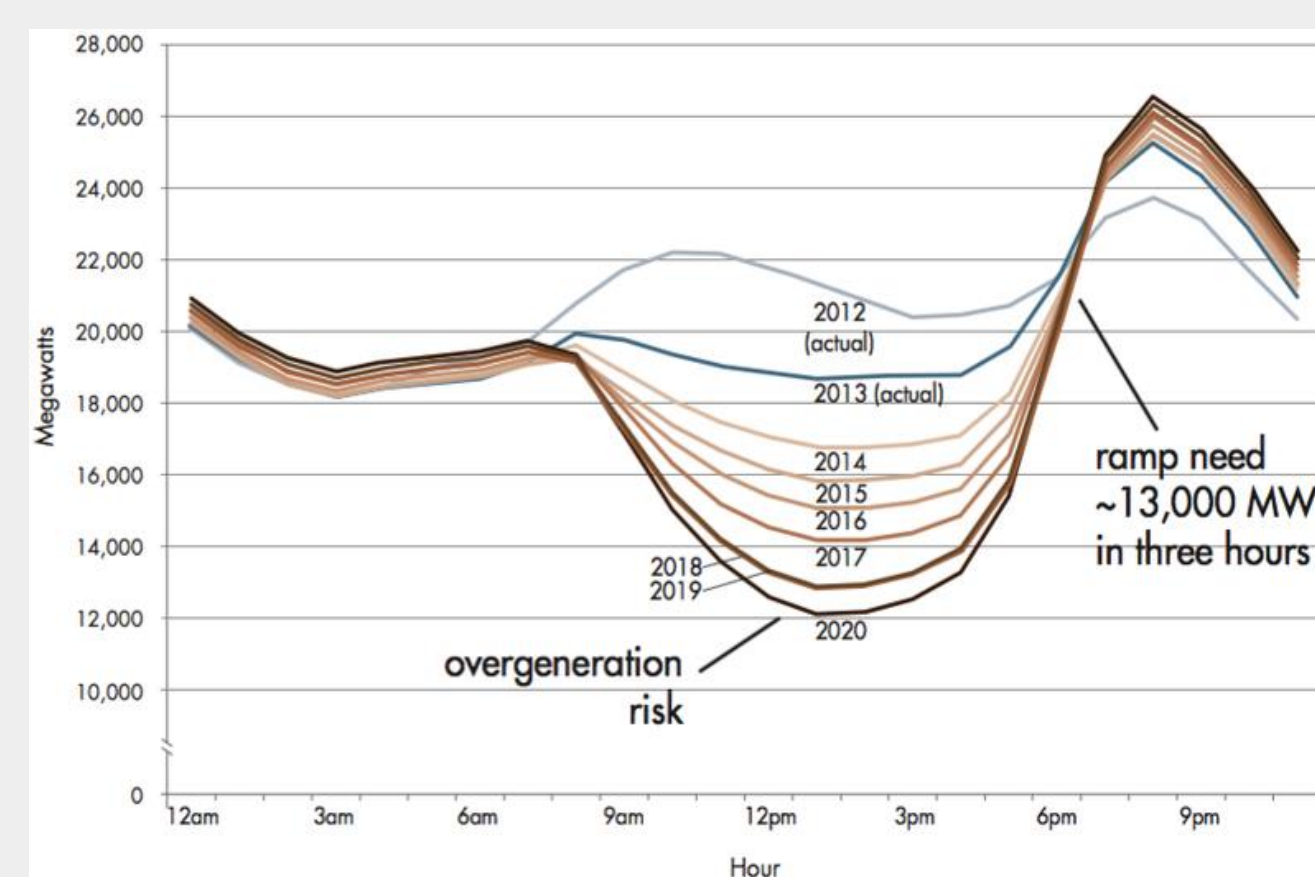


## Objective

**GGGH aligns energy use with renewable energy generation.**

Take energy demands from users and execute them based on when energy generation is cleanest and most readily available

## A Tremendous Problem



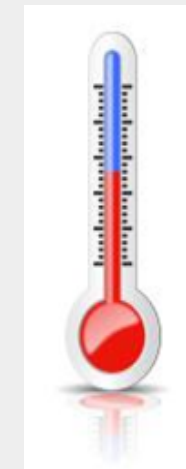
Renewable energy and EV's can cause logistical problems for utilities which require expensive solutions such as peaker plants, storage, and additional distribution maintenance.

## WattTime and Deployable Loads

Python pulls data from the WattTime API (watttime.org), which provides the carbon intensity of electricity coming from the grid on a given day. By controlling shapeable and deferrable loads, energy demands can be met with cleaner energy sources.

**Shapeable loads** can receive a variable amount of power delivered within a flexible timeframe as long as the energy requirement is met by the end of the time frame.

**Deferrable loads** can be scheduled to run at any time within the user defined time frame, but once they begin, they have specific power requirements for a set time.



## Optimization

We used a mixed integer linear program (MILP) to perform the optimization. The cost function is the carbon intensity of the grid (pulled from WattTime); the decision variables are the time and power consumption of the loads; the constraints include car and laundry deadlines, the energy-power relationship, and other appliance-specific constraints.

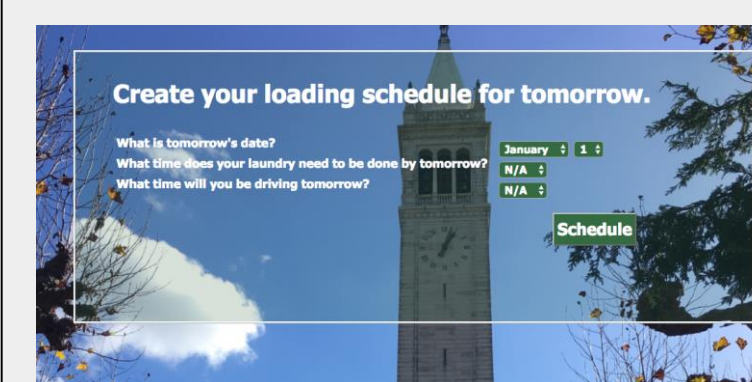
**211** Decision Variables

**633** Constraints

## The Cyber Physical System

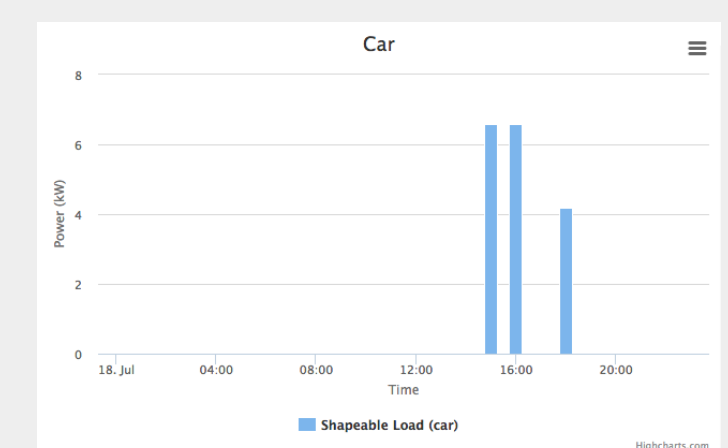
### User input:

User limits time frame in which loads should be complete.



### Visualization:

Charts to summarize load schedules and emissions from energy use.



**WattTime:** provides real time and predicted data about the carbon footprint of electricity coming from the grid



### Wallflower:

Stores and retrieves time series data

### Python:

Provides a central hub for receiving, analyzing, and distributing data.

### Optimization:

Optimization problem that creates load schedules to minimize carbon emissions of electricity use

### Arduino:

Simulated scenario where loads are actuated for a single day in a model home.



## Results

**20%**

Reduction in CO<sub>2</sub> emissions

Equivalent to 100 gallons of gasoline or  
3,000 miles traveled

**1,825 lbs**

Potential yearly CO<sub>2</sub>-e saved by  
using optimization

Implementation in 160,000 homes would prevent the need for  
construction of a \$600 million Natural Gas power plant

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