# Re-Use of Electric Vehicle Batteries for Smart Grid Energy Storage Systems

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#### 1. Societal Need

Energy is a necessity and there is a projected global reliance on fossil fuels in 2035.

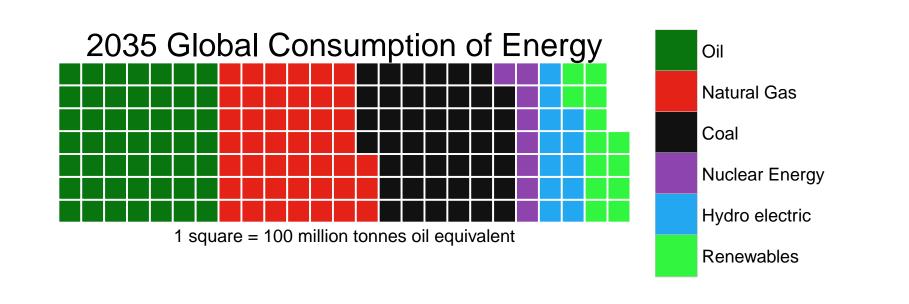


Figure 1: 2035 Global Projected Energy Consumption[1]

CO<sub>2</sub> from **burning fossil fuels** leads to a green-house gas (GHG) effect. Scientific evidence shows causality (through complicated correlation) between **GHG emissions** & global warming.

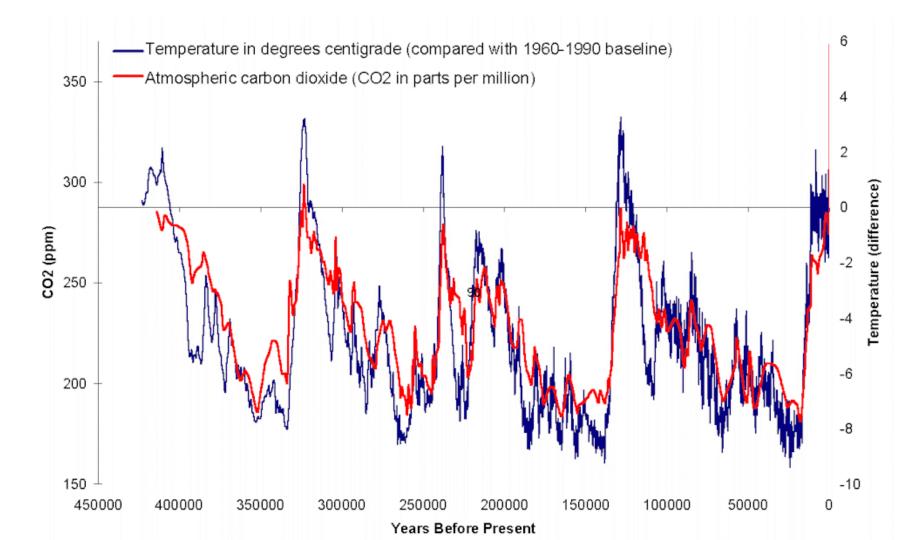
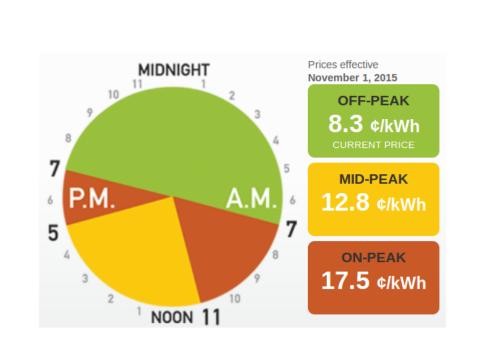


Figure 2: Observed Temperature and CO<sub>2</sub> Ice Core Records[2]

## 2. Problem

Utility companies rely on fossil fuels for peak demand energy generation since most renewable energy is intermittent and nuclear power is constant. The current **peak shaving** solution (i.e. time of use pricing) does not create true **load leveling**.



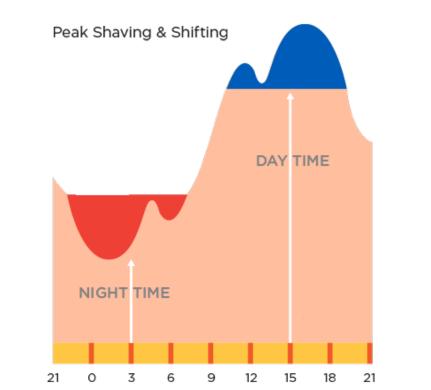


Figure 3: Ontario Time of Use Electricity Pricing[3]

Figure 4: Peak Shaving

Also the increasing number of plug-in electric vehicles (EV) produces new **battery waste**.

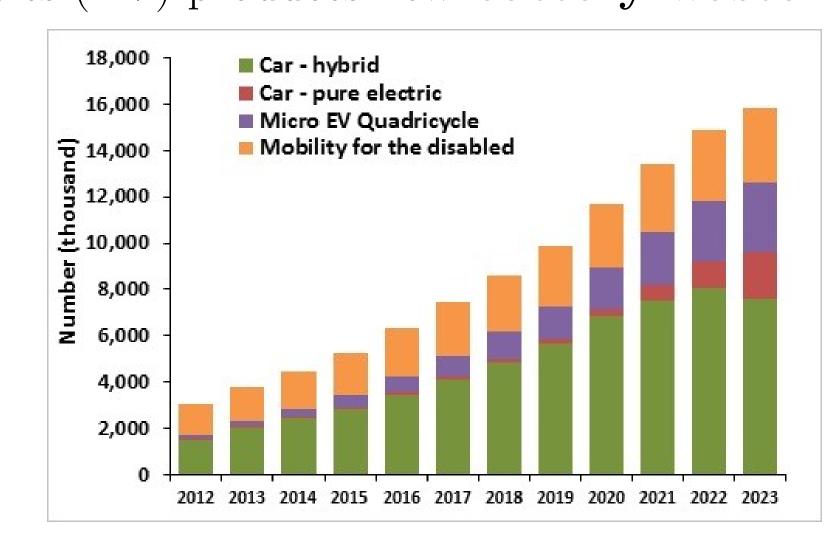
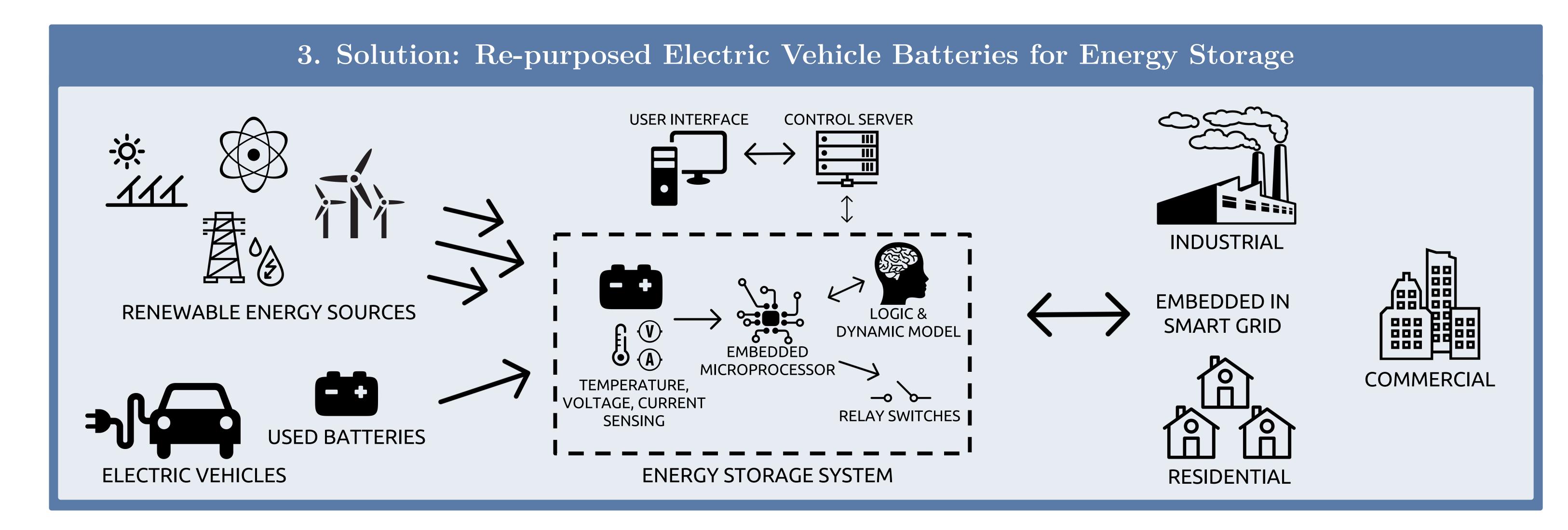


Figure 5: Forecasted EV sales [4]



## 3.1 Battery Modeling

EV batteries are disposed when they cannot meet the vehicle voltage requirements. Research shows that after an irreversible capacity fade, disposed batteries can still hold up to 80% of charge.

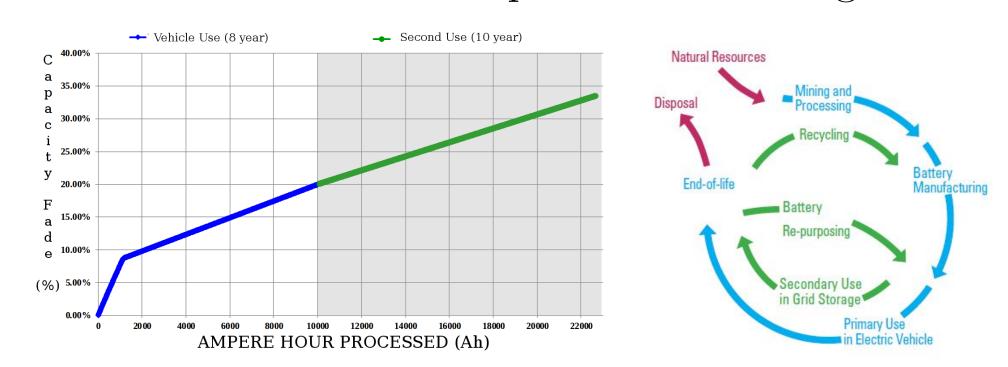


Figure 6: Capacity Fade on Figure 7: Proposed Battery[5] Battery Lifecycle[6]

The first step was to create and verify a system model of the re-purposed batteries.

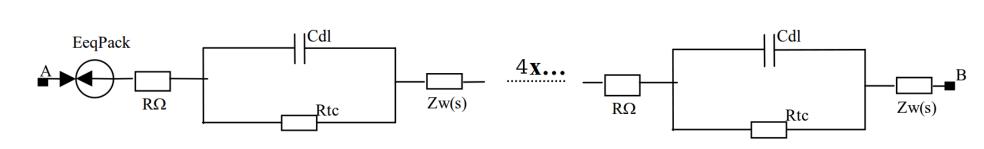


Figure 8: Battery Pack (Plant) Model

System Equations and Data Validation: Voltage Response of Given Current:

$$v_{eqpack}(t) = E_{pack} + R_{\Omega}i(t)$$
  
+4  $\left[\mathcal{L}^{-1}\left(\frac{R_{tc}I(s)}{1+sR_{tc}C_{dl}}\right) + \mathcal{L}^{-1}\left(\frac{(1+s\tau_2)^{n_2}I(s)}{(s\tau_1)^{n_1}}\right)\right]$   
**Energy Flow** in Battery System:

Energy Flow in Battery System:  $e(t) = \int_{t=0}^{t} p(t)dt = \int_{t=0}^{t} \left[ E_{eqpack} + 6 \left( R_{\Omega} j(t) + \sum_{n=1}^{15} v_n(t) \right) \right] i(t)dt$ Heat Generation (HG) in Battery System:  $q = \dot{Q} = \int \delta Q/t = Q_{charged \to discharged}/t$ 

 $q = (I(U-V) + I(T\frac{dU}{dT}))/t$ 

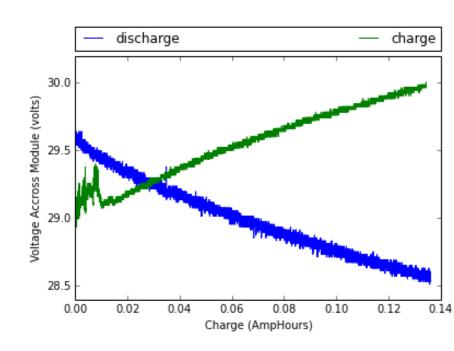


Figure 9: State of Charge (Coulomb Counting)

0.4 0.2 0.0 -0.2 -0.4 -0.6 -0.8 -1.0 500 1000 1500 2000 2500 3000 3500 4000 elasped time (s)

Figure 10: Voltage Response of Battery Cell to Current

## 3.2 Electrical Design

After bench-marking the batteries, an electrical system was designed and implemented to safely monitor and control the charging and discharging.

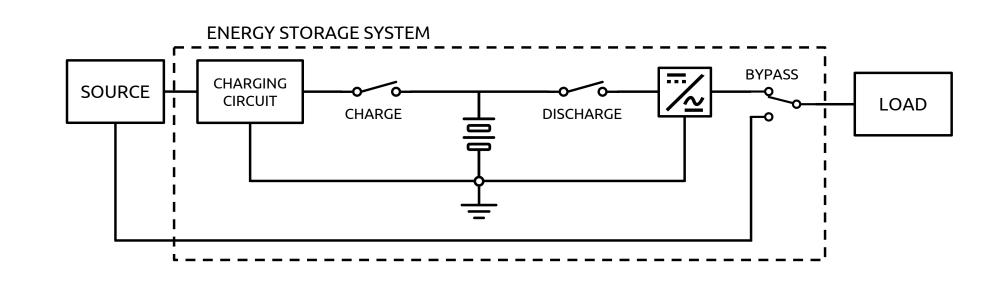


Figure 11: Simplified System Circuit

There were five parallel packs of four 7.2 V, 6.5 Ah Toyota Prius NiMH battery modules in series. Useful Capacity: 480 Wh (see [7])
Nominal Voltage: 28.8 V

## 3.3 Mechanical Design

The enclosure was designed to securely house the batteries, allow for adequate temperature control and meet safety requirements.

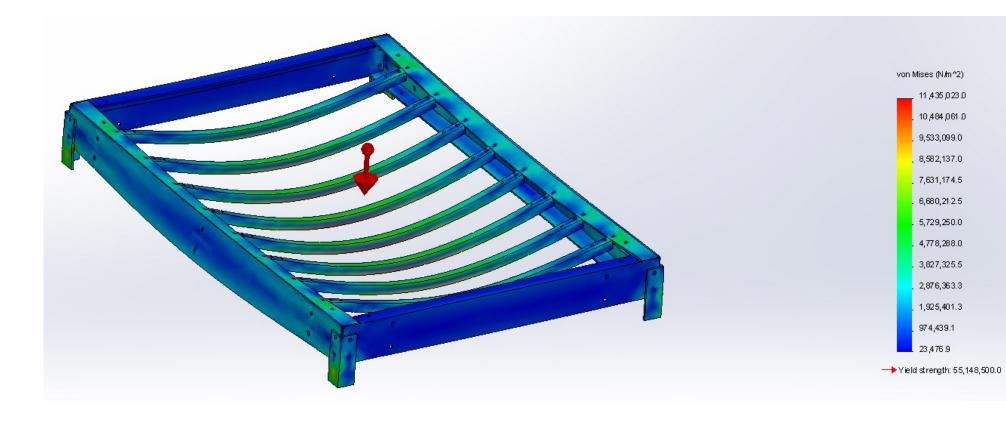


Figure 12: Bottom Frame Stress Analysis

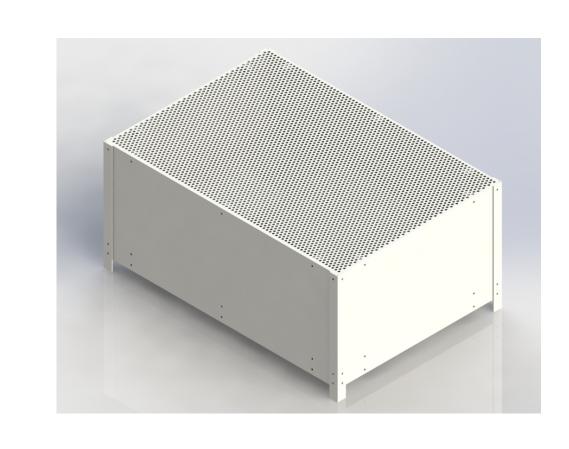


Figure 13: Full 3D Photo-realistic Render

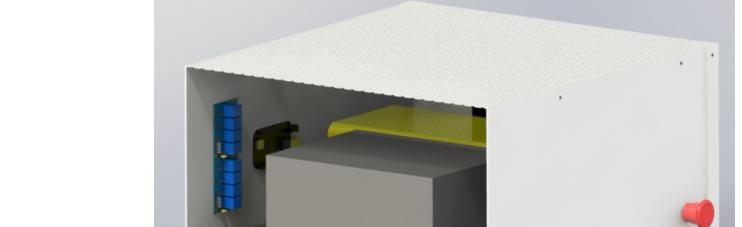


Figure 14: Cross Sectional View

## 3.4 Controls

Next, a closed loop control system was designed to meet specifications and requirements.

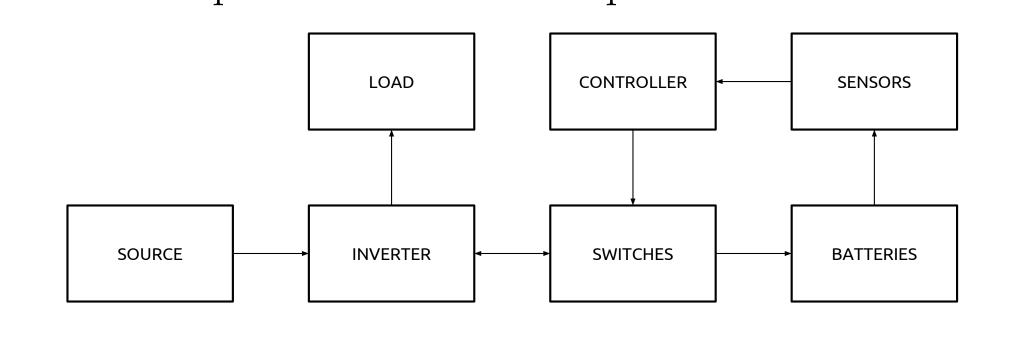


Figure 15: High Level System Block Diagram

Demand Forecasting (DF): Autoregressive
Moving (Winter's) Triple Exponential Smoothing
State of Charge (SOC): Extended Kalman
Filtering paired with Coloumb Counting
Remaining Useful Life (RUL): Support
Vector Machine (non-linear) Charge Regression
Dynamic Temperature Control: Fuzzy
rule-sets based upon HG, DF, SOC and RUL
Internet of Things (IOT):

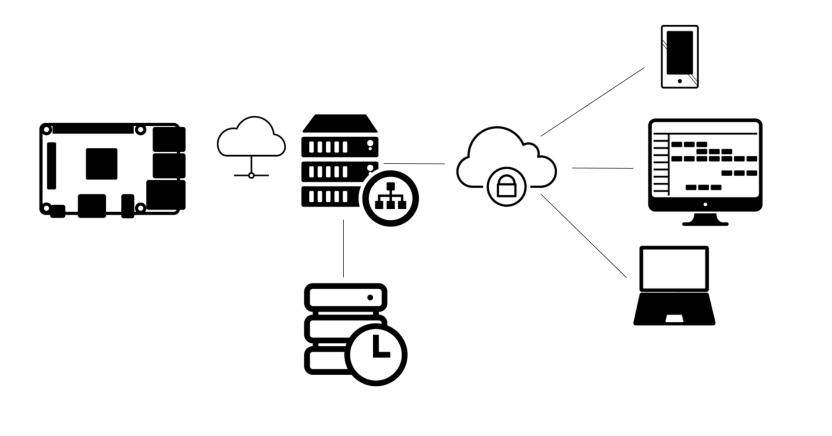
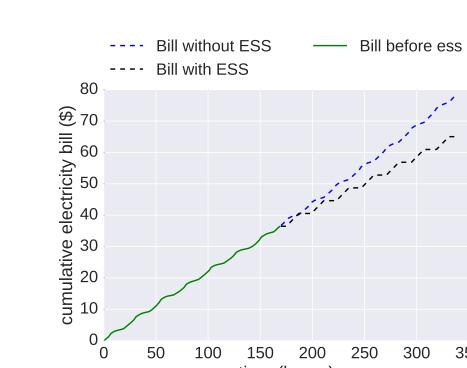


Figure 16: Internet of Things System Diagram

## 4. Results

In one winter month in Ontario, at the grid level, stationary energy storage can replace 788 GWh of demand normally supplied by natural gas, eliminating 143 Mt of CO<sub>2</sub> emissions[8]. A household may save \$50 on their monthly electricity costs.



18000 fossil free supply charging charging 16000 15000 12000 20 40 60 80 100 120 12 time (hours)

Figure 17: Monthly Savings with Energy Storage

Figure 18: Grid Level Supply and Demand with Energy Storage

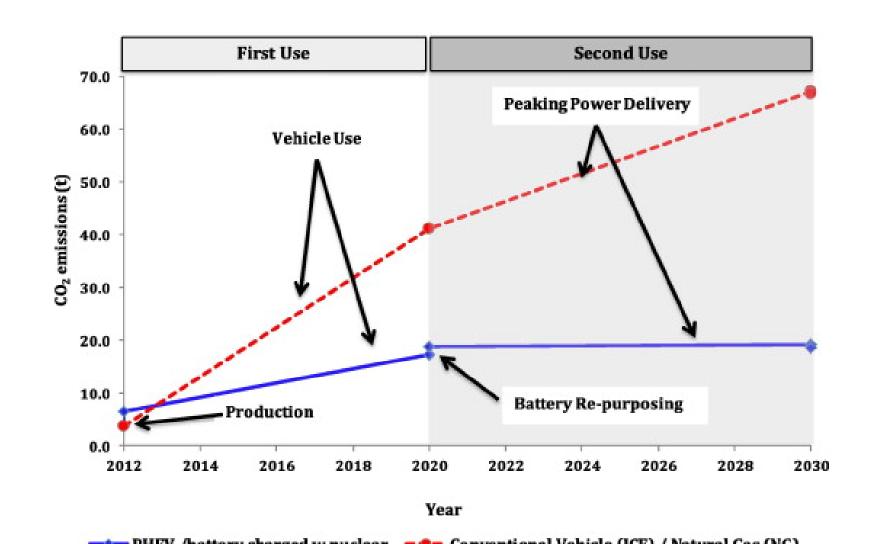


Figure 19: Simulation of CO<sub>2</sub> Emissions [6]

## 5. Conclusion

The prototype built successfully demonstrated the ability of re-purposing EV batteries for distributed stationary energy storage to create true load leveling in the energy grid.

#### References

- [1] April 2015 BP statistical review of world energy.
- [2] NOAA Paleoclimatology Program.
- Temperature change and carbon dioxide change.
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- Hybrid & pure electric vehicles for land, water & air 2013-2023: Forecasts, technologies, players.
- [5] Leila Ahmadi et al.
- Energy efficiency of li-ion battery packs re-used in stationary power applications.
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- [8] EIA Carbon Dioxide Emissions Coefficients.
- [9] Sean B. Walker et al.
- Incentives for the reuse of electric vehicle batteries for load-shifting.