

Coupled Thermal- Electrical Modeling of Integrated Photovoltaic Systems

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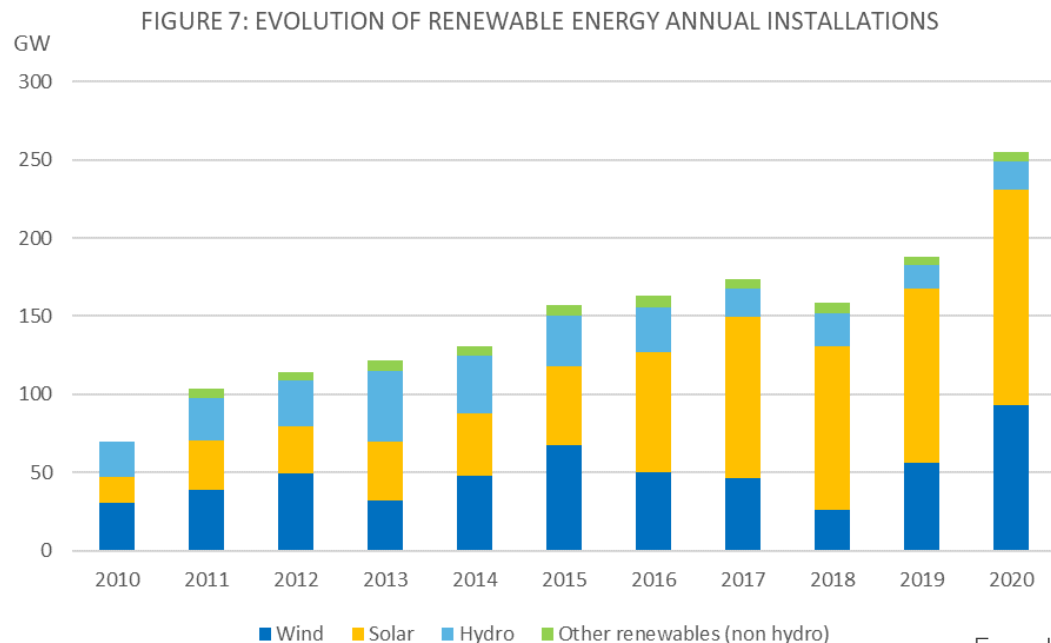
Outline

- Introduction
- Methods
- Examples
 - Residential Rooftop PV
 - Evaluating the Energy Production of Vehicle-Integrated Photovoltaics (VIPVs)
 - Electric Vehicle Energy Consumption with On-board PVs

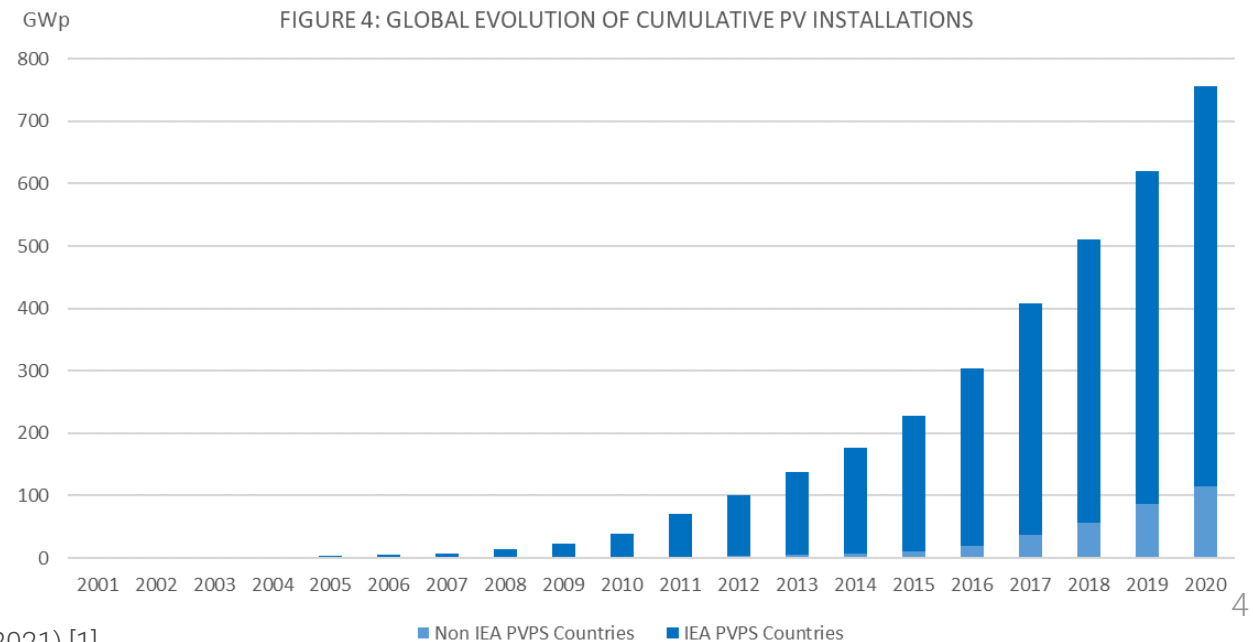
Introduction

PV Industry Background

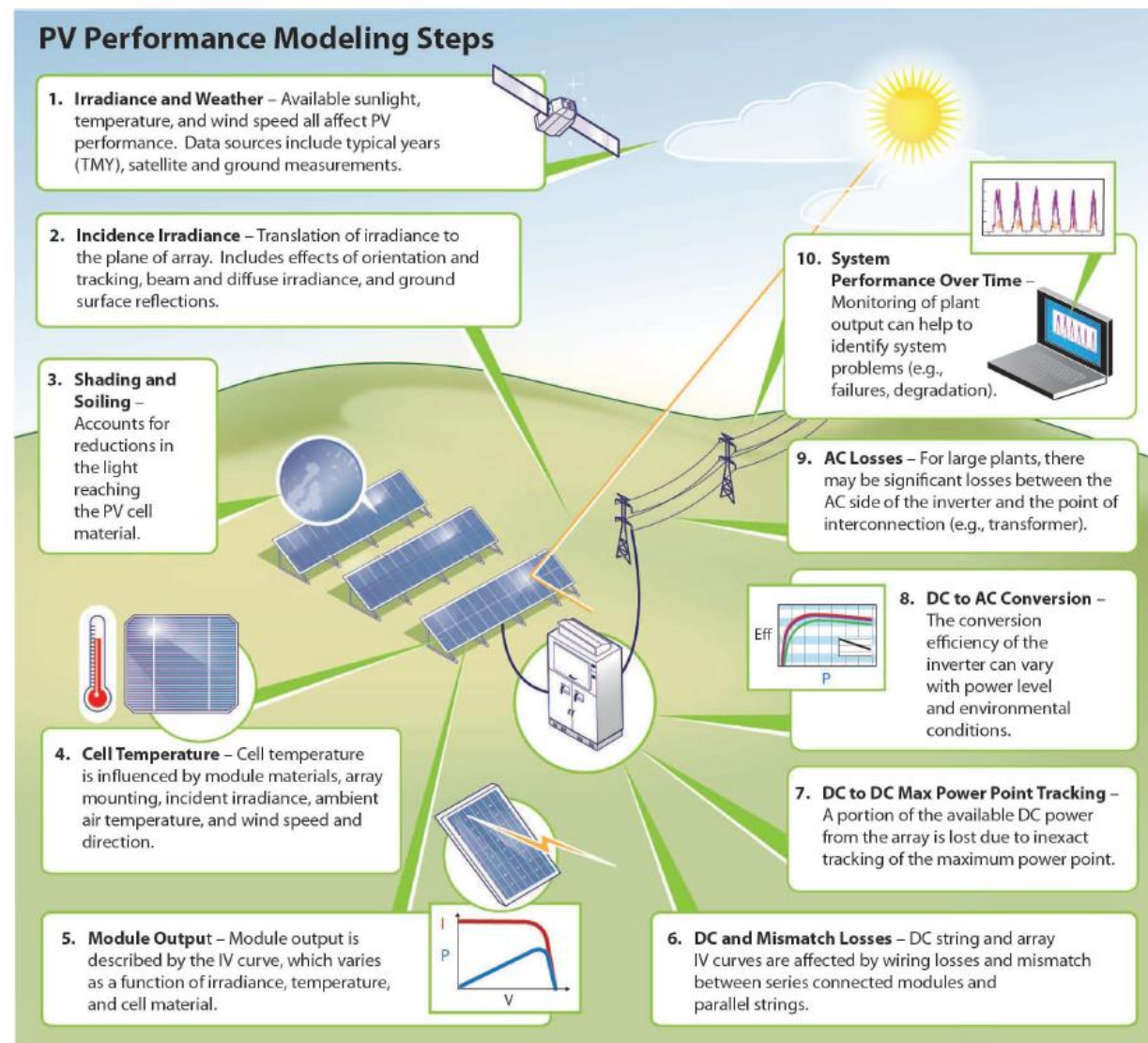
- Solar power the fastest growing renewable energy source
- U.S. solar power capacity has experienced an average annual growth rate of 33% in the last decade [1]
- Global market of \$160 billion in 2021 [2]



From IEA (2021) [1]

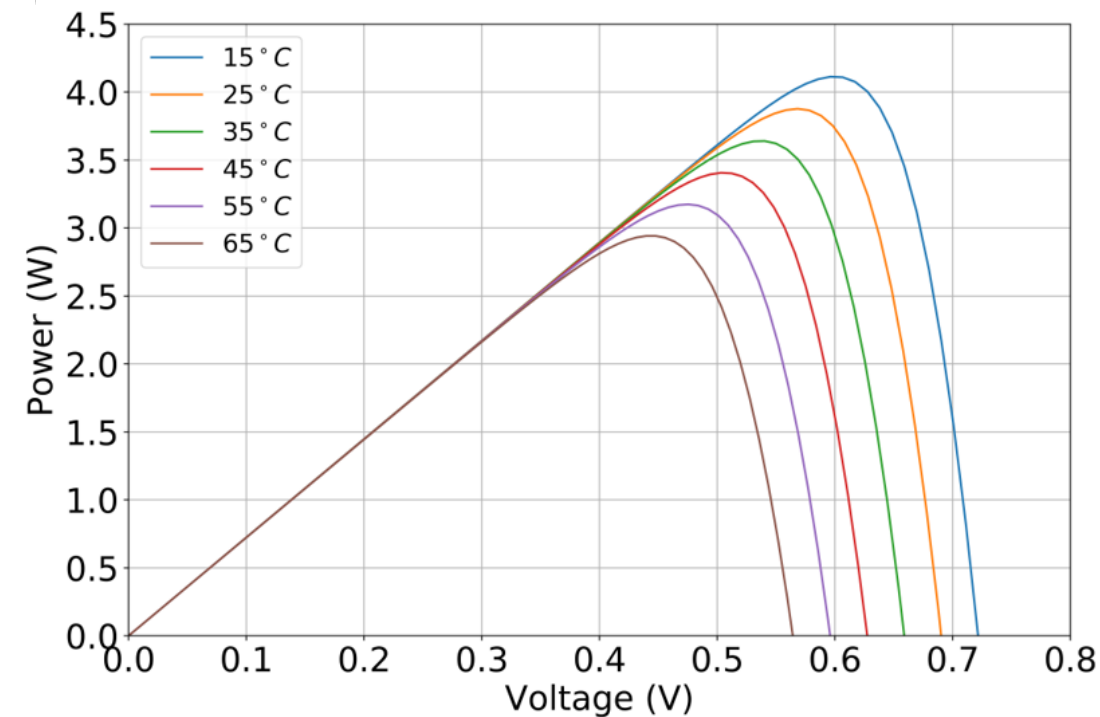


PV System Modeling



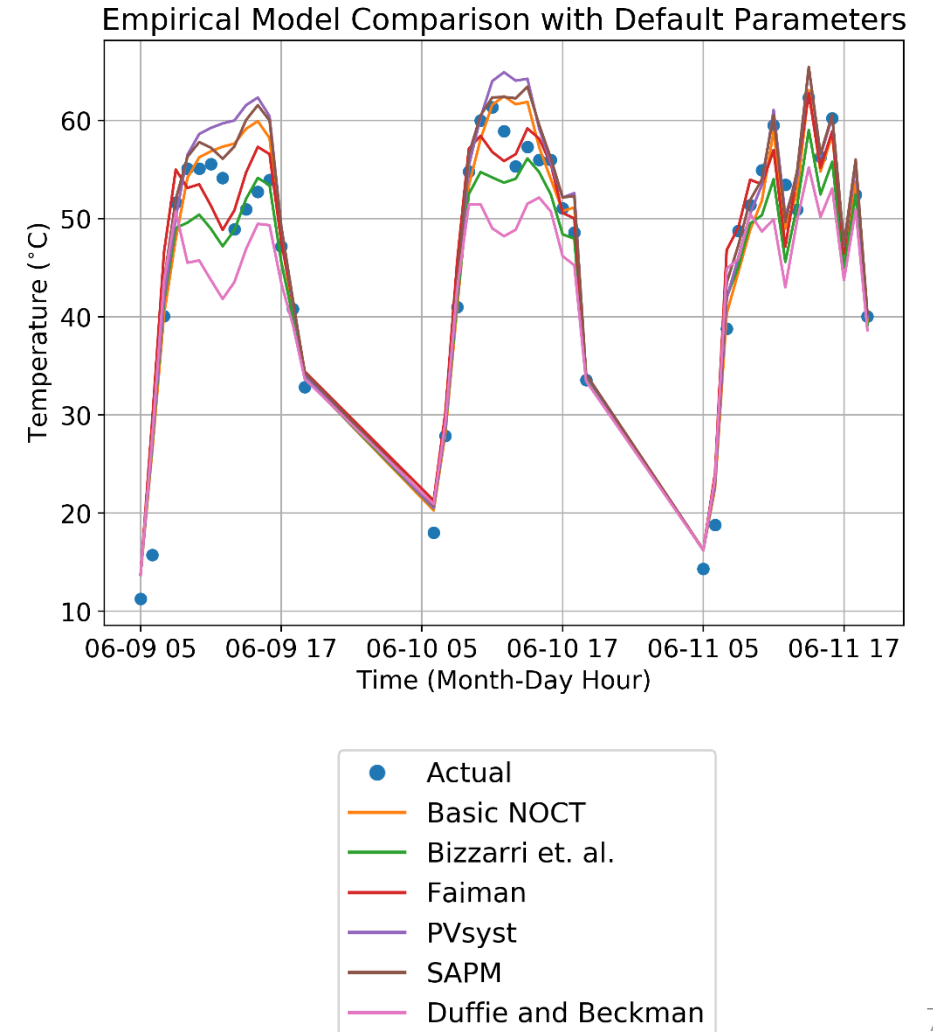
Why Consider PV Temperatures?

- Solar cell efficiency typically decreases with increasing temperature
- Performance is often rated at 25 °C, while cells in operation reach significantly higher temperatures
- Therefore, cell T is needed for accurate energy production prediction



Limitations of Empirical Cell Temperature Models

- Many PV simulation libraries/software include empirical models for cell temperature
- Often these are used with default coefficients which are provided by the library based on literature
- Large disagreement between models



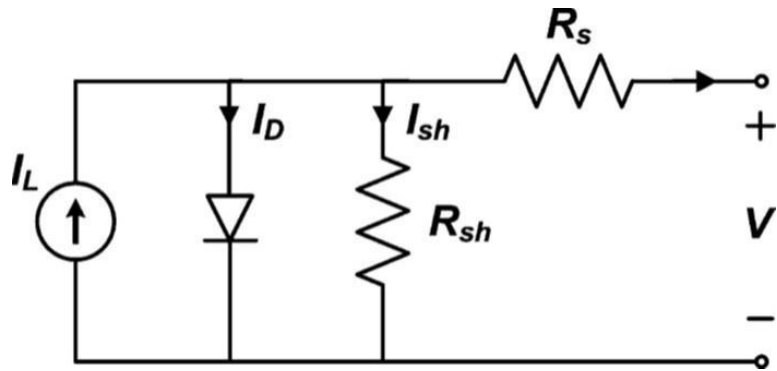
Advantages of Heat-Transfer Solver for PV Modeling

- Improved accuracy of temperature predictions for solar modules, especially if that are integrated into structures (e.g., vehicles and buildings)
- Analysis of the impact of PVs on the structures' temperatures
 - E.g.: How could cabin T and occupant comfort change if PV films are integrated into building windows or vehicle sunroof?
- Design of PV cooling techniques

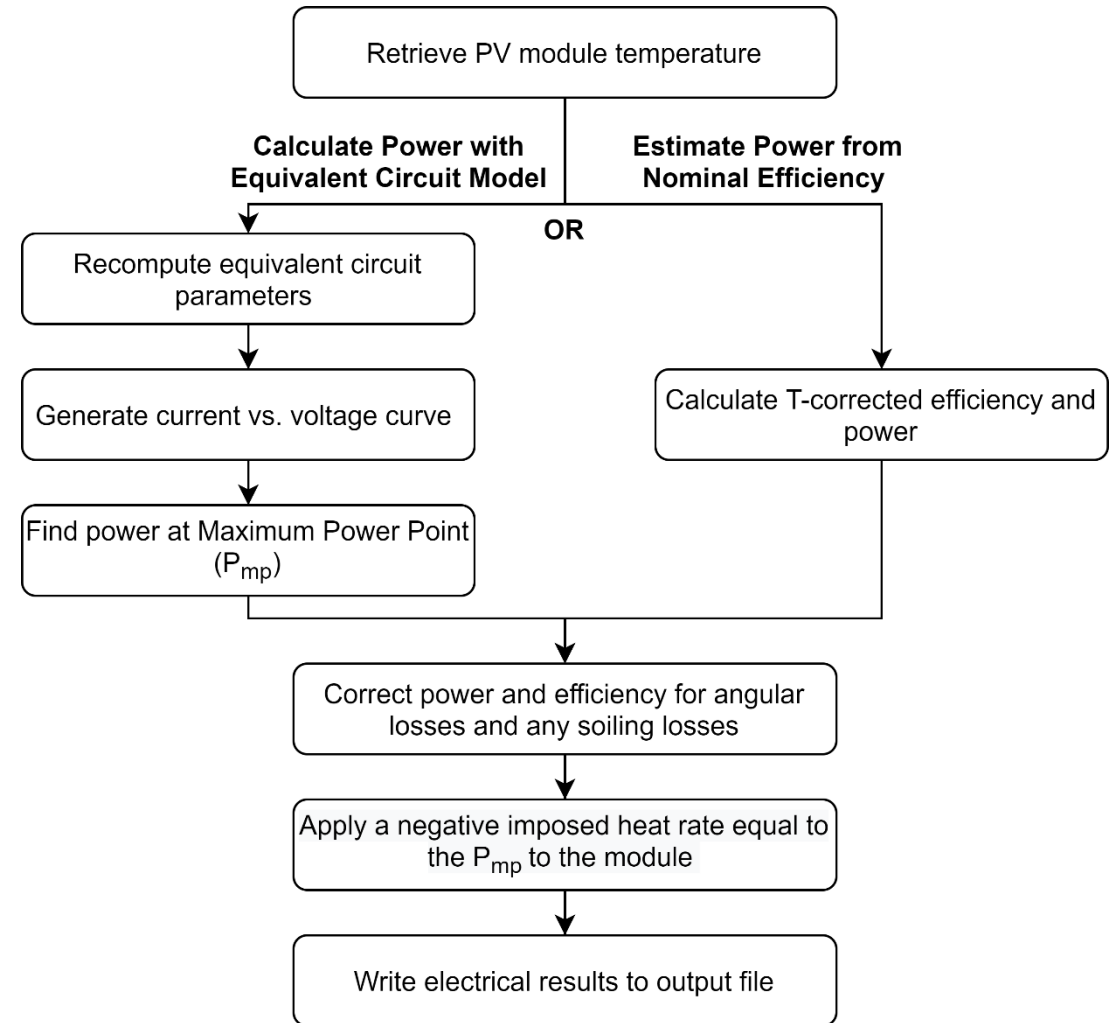
PV Modeling with TAlTherm

Coupled Thermal-Electrical Model

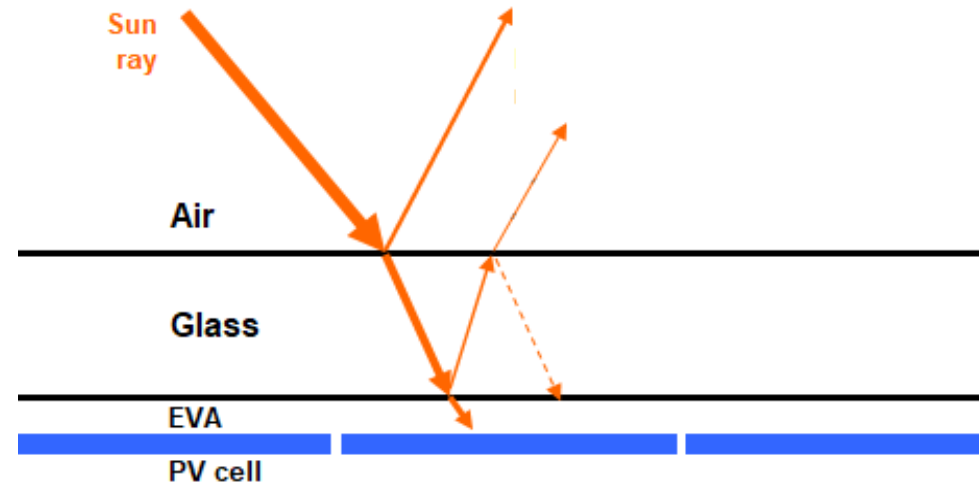
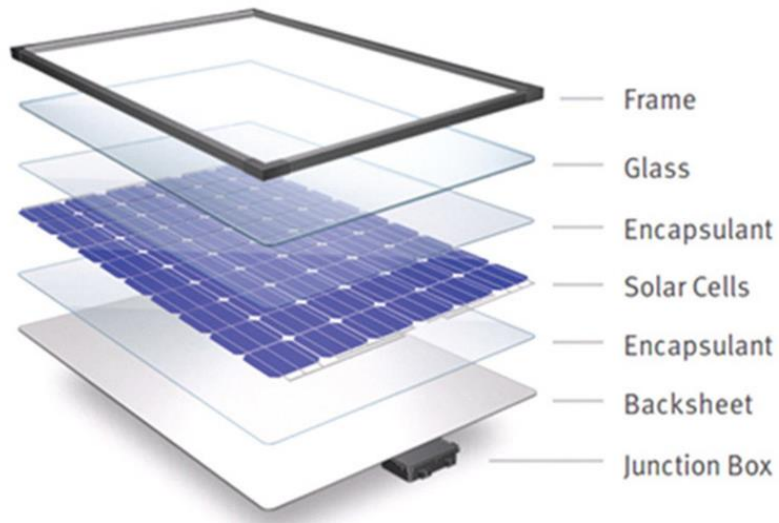
- PV electrical equations solved at end of each thermal simulation time-step
- Model inputs use information available from manufacturer datasheets



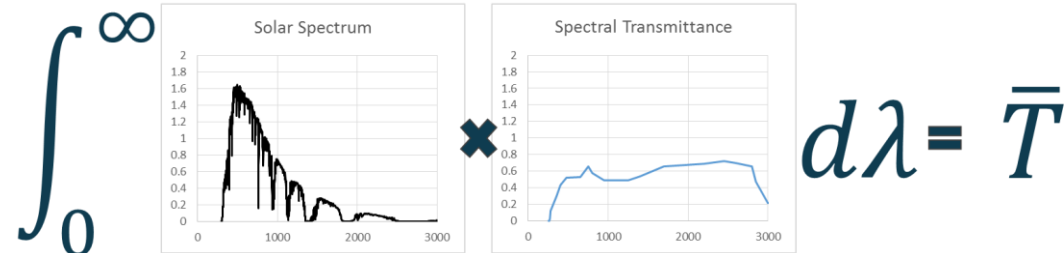
$$I(V) = I_L - I_0(e^{(V+IR_s)/a} - 1) - \frac{V + IR_s}{R_{sh}}$$



TAItherm Transparent Materials Modeling



- Specify transmittance and reflectance for each layer as band-averaged values or spectral curves in TAItherm
 - Curves will be band-averaged based on the solar spectrum



PV Modeling Extension Prototype Demo

Photovoltaics Modeling Extension

File Help

Main Options | Electrical Params | Optical Params | System Losses

PV Model Name
Jeep_w_solar_roof_hood

PV Part Names (comma seperated)
Roof.Hood

☐ Perform Element Level Calculation

☐ Write All Element Results

Cancel Save and Exit Save and Run

Photovoltaics Modeling Extension

File Help

Main Options | Electrical Params | Optical Params | System Losses

Approach Equivalent Circuit

Equivalent Circuit

Derive Parameters

α_s 0.000789 A/C

a (ref) 2.9703606 V

IL (ref) 1.839585 A

Is (ref) 3.9159574e-13 A

R_s (ref) 5.8283625 Ohms

R_{sh} (ref) 541.61916 Ohms

Adjust -14.33575 %

Method Lambert W

Calculate IV Curves

Efficiency Estimate

Efficiency %

Temperature Coefficient %/C

Scale Electrical Values to TAItherm Part/Element Areas

Module Area m²

Cancel Save and Exit Save and Run

EqC Parameter Derivation

Cell Type CdTe

Voltage at Pmax (Vmpp) 67.8 V

Current at Pmax (Impp) 1.62 A

Open Circuit Voltage (Voc) 86.4 V

Short Circuit Current (Isc) 1.82 A

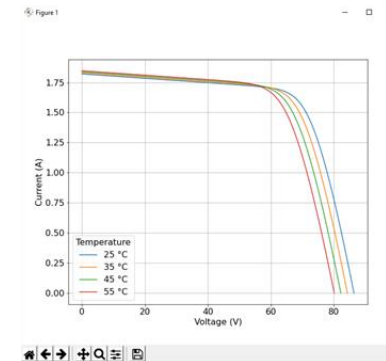
Temperature Coeff of Isc 0.000789 A/C

Temperature Coeff of Voc -0.24192 V/C

Temperature Coeff of Pmax -0.28 %/C

Cells in Series 216.0

Derive Equivalent Circuit Parameters



Photovoltaics Modeling Extension

File Help

Main Options | Electrical Params | Optical Params | System Losses

Approach IAM Polynomial

IAM Polynomial

Specify coefficients of a 5th order polynomial curve-fit

a -5.315717e-10

b 6.266005e-8

c -3.258396e-6

d 4.90089e-5

e 0.0001102294

f 0.9965551

Plot IAM

Martin and Ruiz IAM Model

Angular Losses Coefficient 0.16

Plot IAM

Use Angular Losses Curve

Angular Losses Coefficients Curve Filename C:/Users/tg/Documents/ Browse

Plot Coeffs

Soiling Losses

☐ Consider Soiling Losses

Soiling Losses Curve Filename C:/Users/tg/Documents/ Browse

Plot Losses

Cancel Save and Exit Save and Run

Photovoltaics Modeling Extension

File Help

Main Options | Electrical Params | Optical Params | System Losses

Mismatch 0.0 %

Wiring 0.0 %

Connections 0.0 %

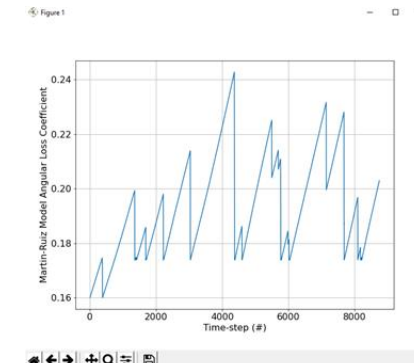
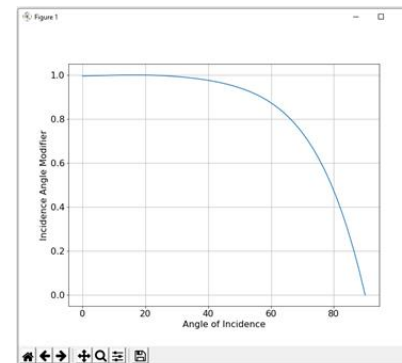
Light-Induced Degradation 0.0 %

Nameplate Rating 0.0 %

Age 0.0 %

Availability 0.0 %

Cancel Save and Exit Save and Run



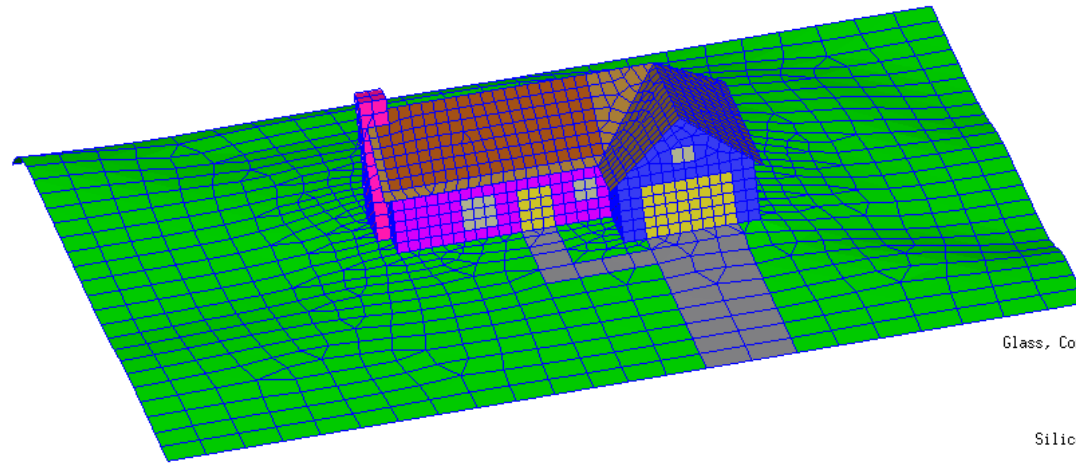
Example Application: Residential Rooftop PVs

Example Residential Rooftop PV Model

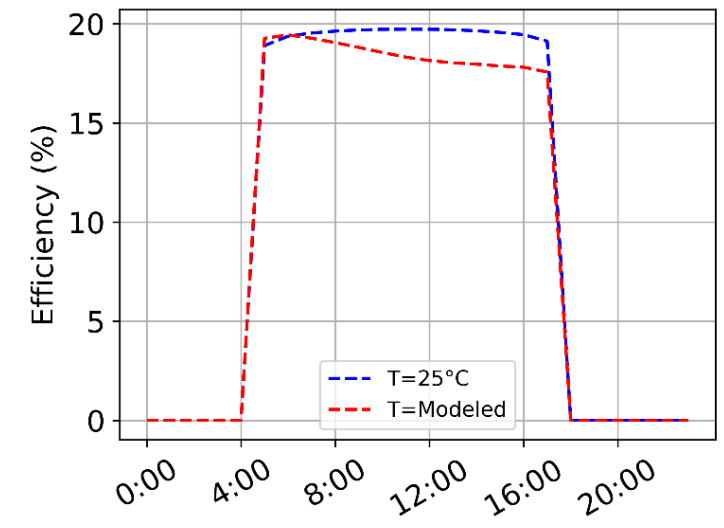
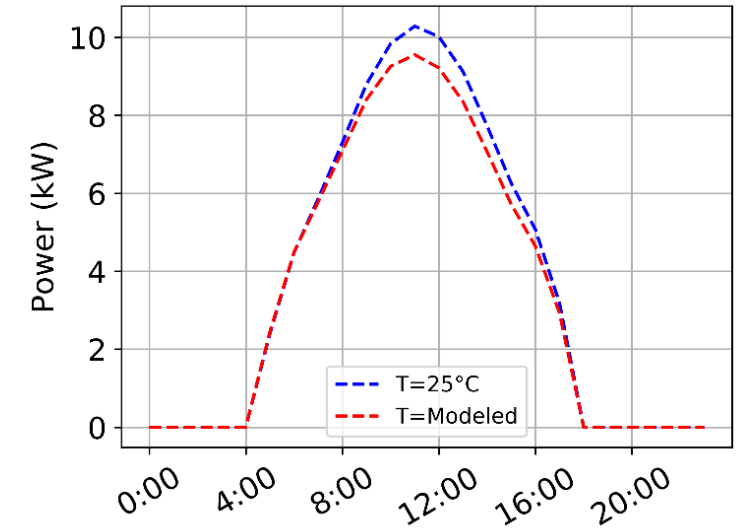
Model Size (mm):
X = 37982
Y = 37982
Z = 8056.2

Visible Size (mm):
X = 37982
Y = 37982
Z = 8056.2

Visible Counts:
Parts = 26
Elements = 3300



Asphalt
Brick
Carpet, 16oz.
Concrete
Glass, Conventional Automotive
Gypsum
PV Glass
Plywood
Silicon Carbide (Temp Dep)
Steel (mild)
Steel, Mild
Wood, Oak
Not Applicable

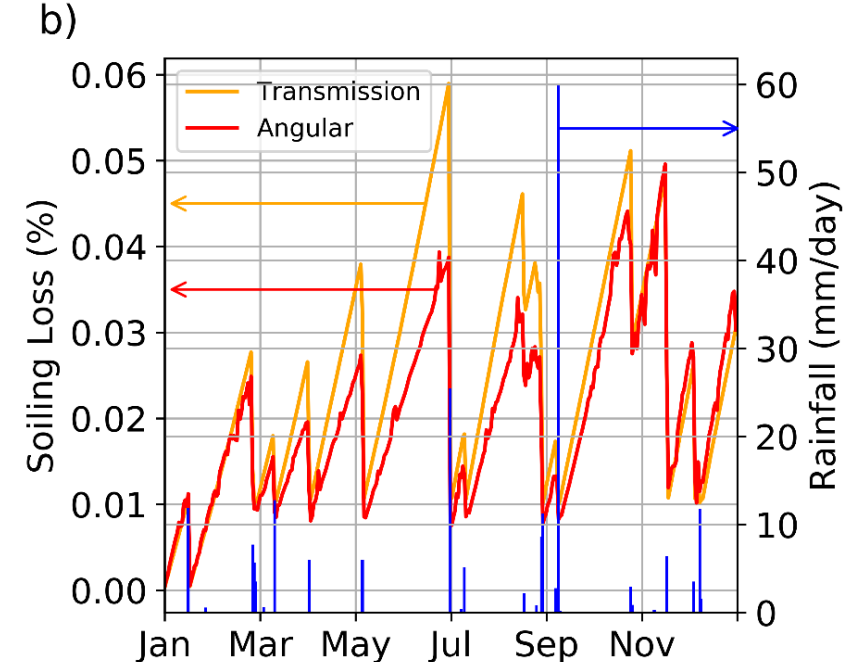
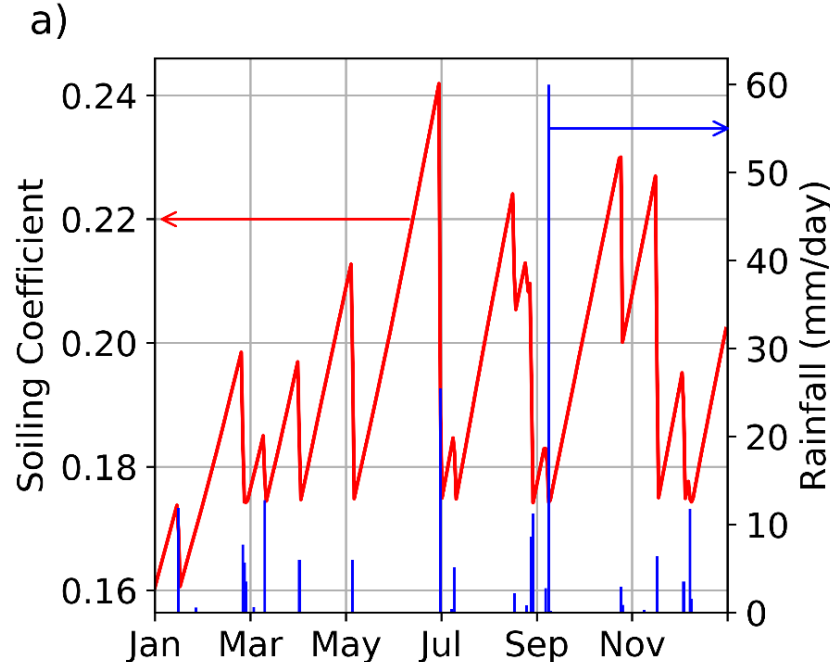


Residential Rooftop PV Model: Considering Soiling Losses

- Soiling loss can be split into a transmission and angular loss contribution

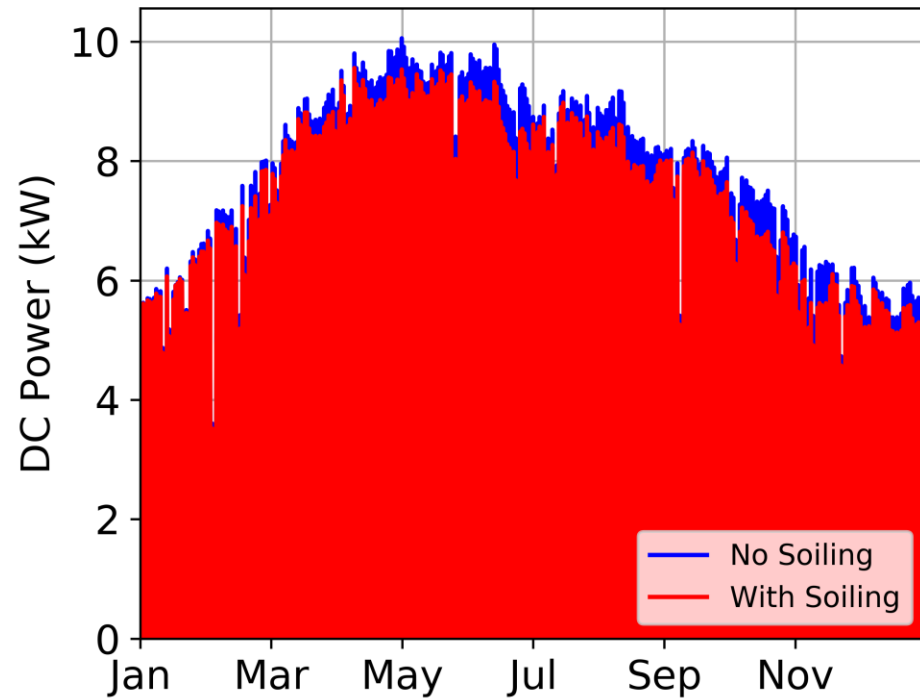
$$L_{trans}^{soiling} = c(a_r^{dirty} - a_r^{clean})/100$$

$$L_{tot} = L_{trans}^{soiling}(1 - K_{\alpha})$$

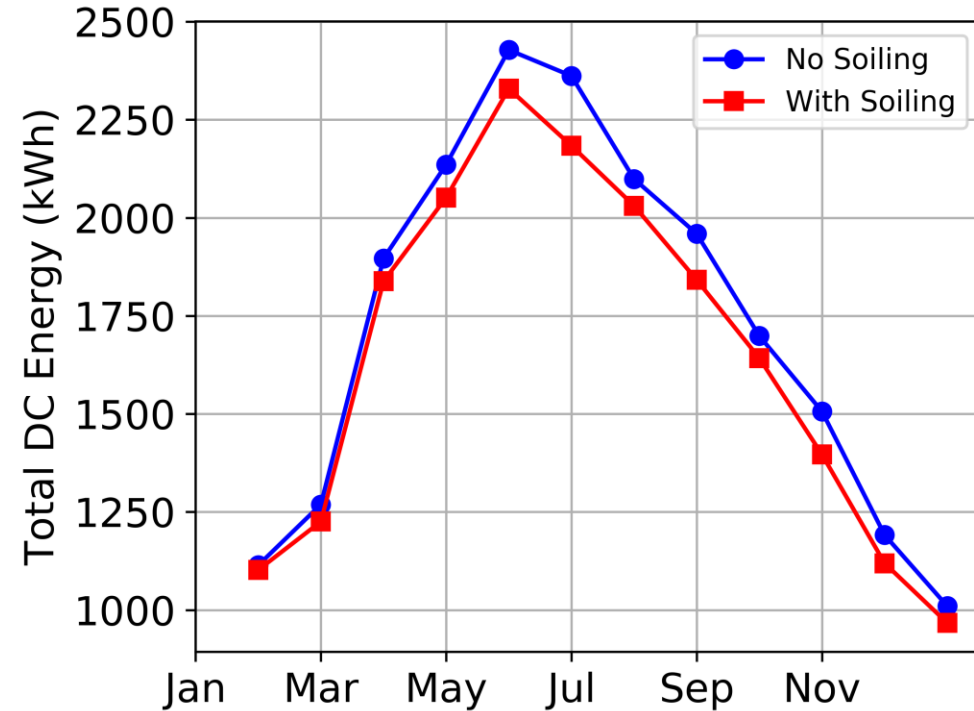


Residential Rooftop PV Model: Considering Soiling Losses

a)



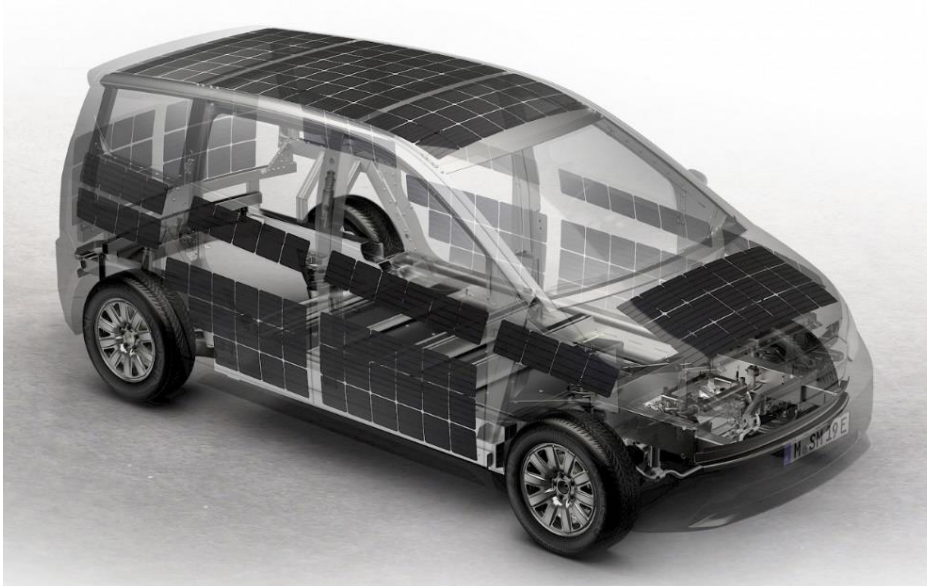
b)



Example Application: Evaluating the Energy Production of Vehicle- Integrated Photovoltaics (VIPVs)

Vehicle-Integrated Solar Panels

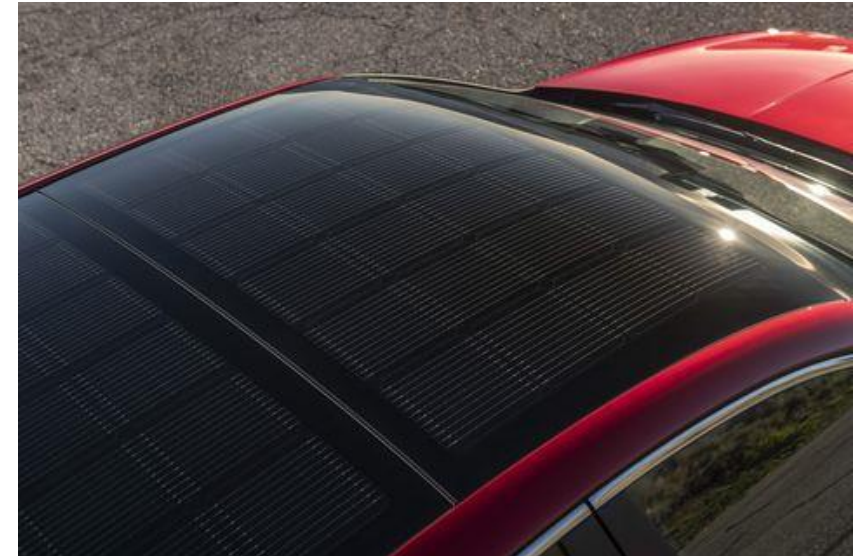
- Extend the range of electric vehicles
- Power an auxiliary battery for secondary electronics systems in both internal combustion engine and electric vehicles



Sono Motors. Up to 34 km/day. Expected 2022. €25,500



Hyundai (2018). Translucent solar roof engineering.



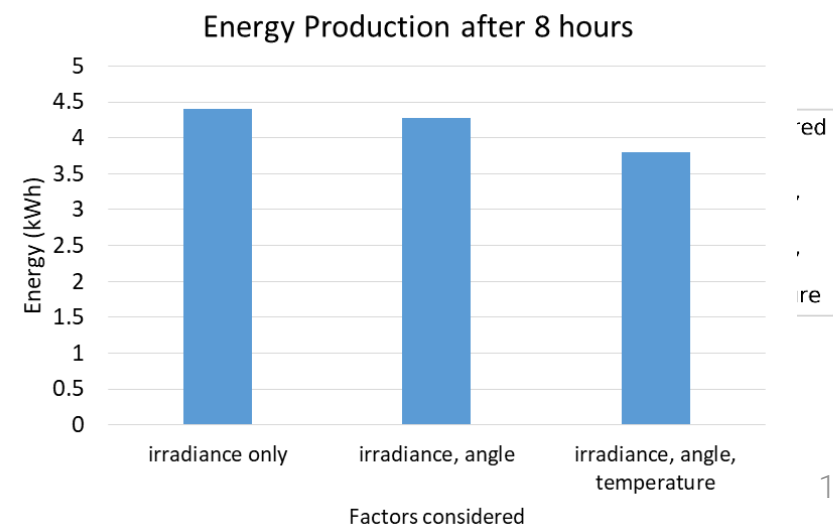
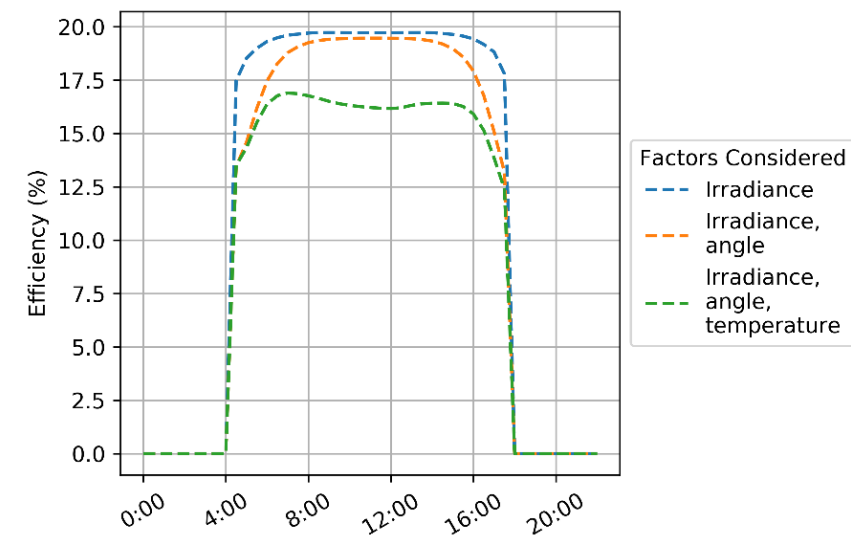
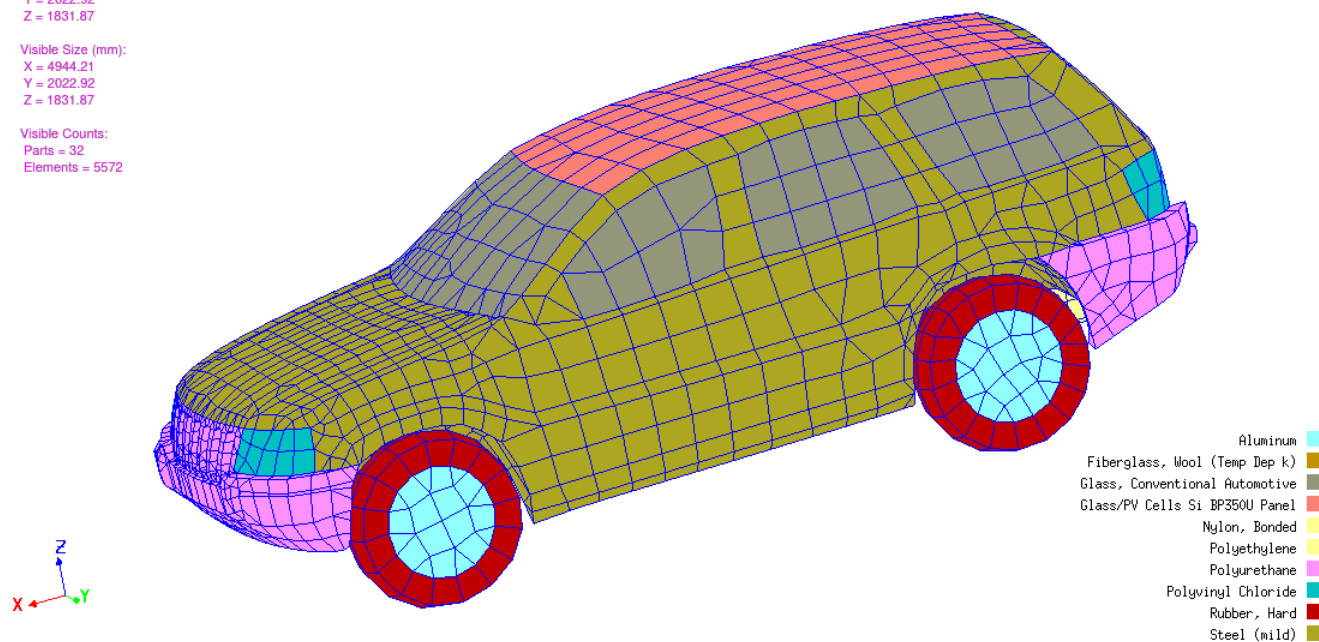
2020 Hyundai Sonata Hybrid

Vehicle-Integrated Solar Panels

Model Size (mm):
X = 4944.21
Y = 2022.92
Z = 1831.87

Visible Size (mm):
X = 4944.21
Y = 2022.92
Z = 1831.87

Visible Counts:
Parts = 32
Elements = 5572



- Predicted 3.8 kWh of energy production per 8 hours in the sun in Phoenix, AZ in July
- 3% angular losses and 13% thermal losses

Wavelength-Selective Transparent PVs

- Absorb UV/IR wavelengths, while transmitting visible light

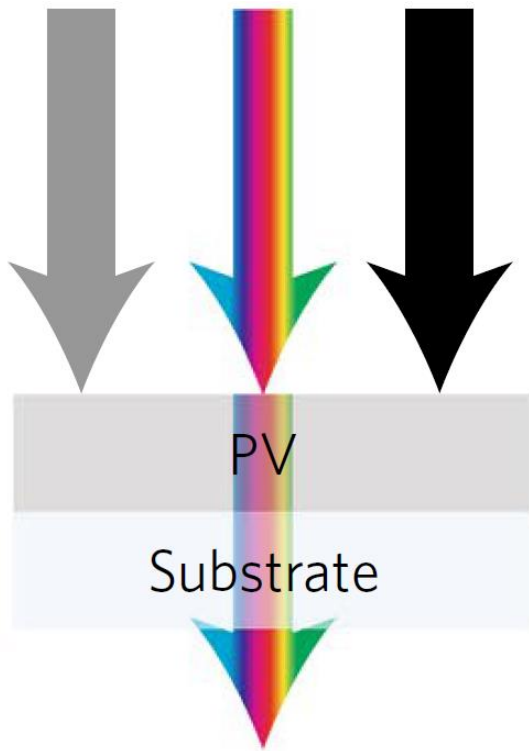
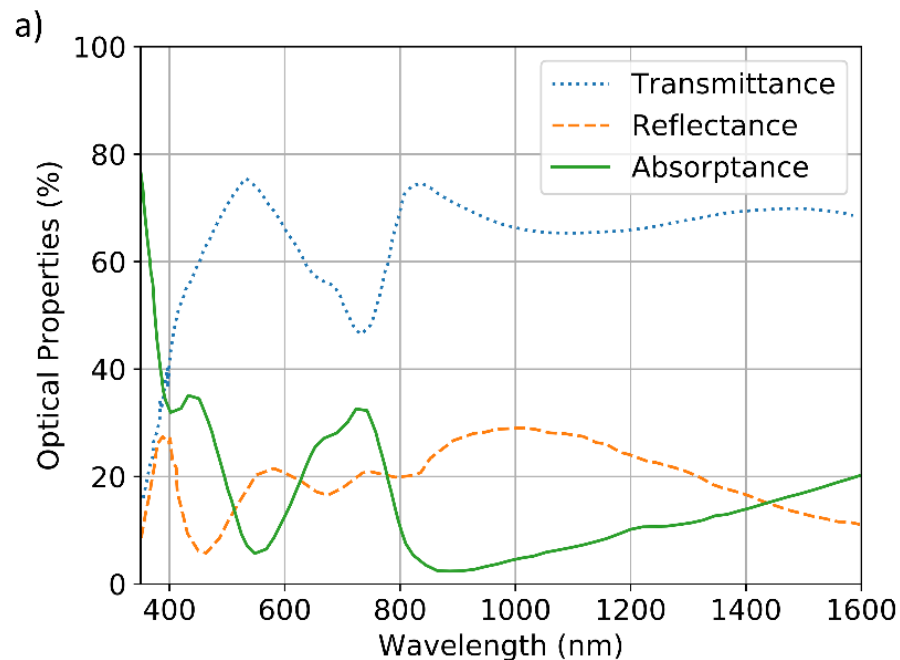
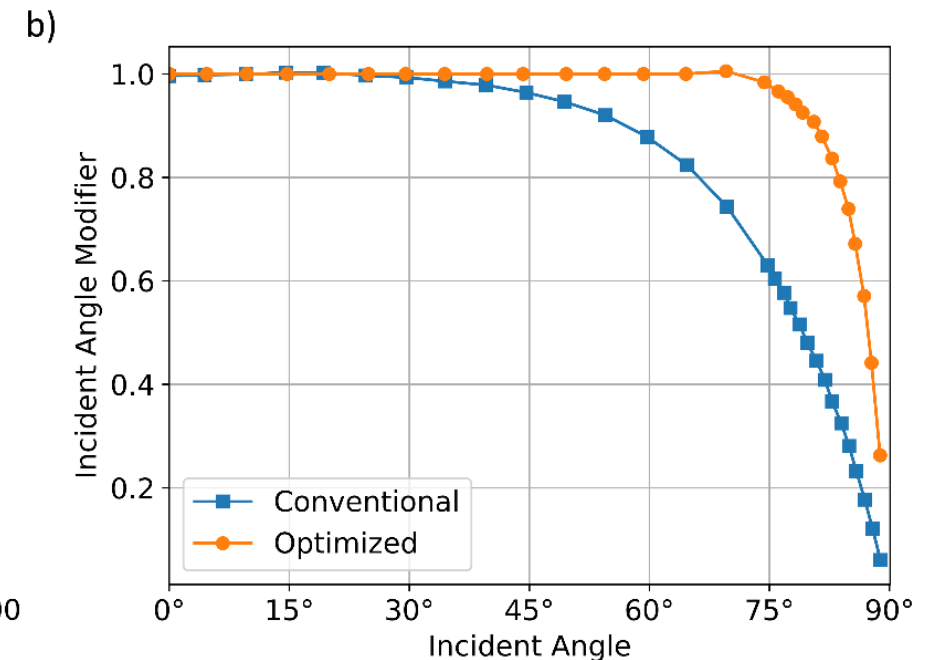


Fig. from Traverse et al. Nature Energy, 2017



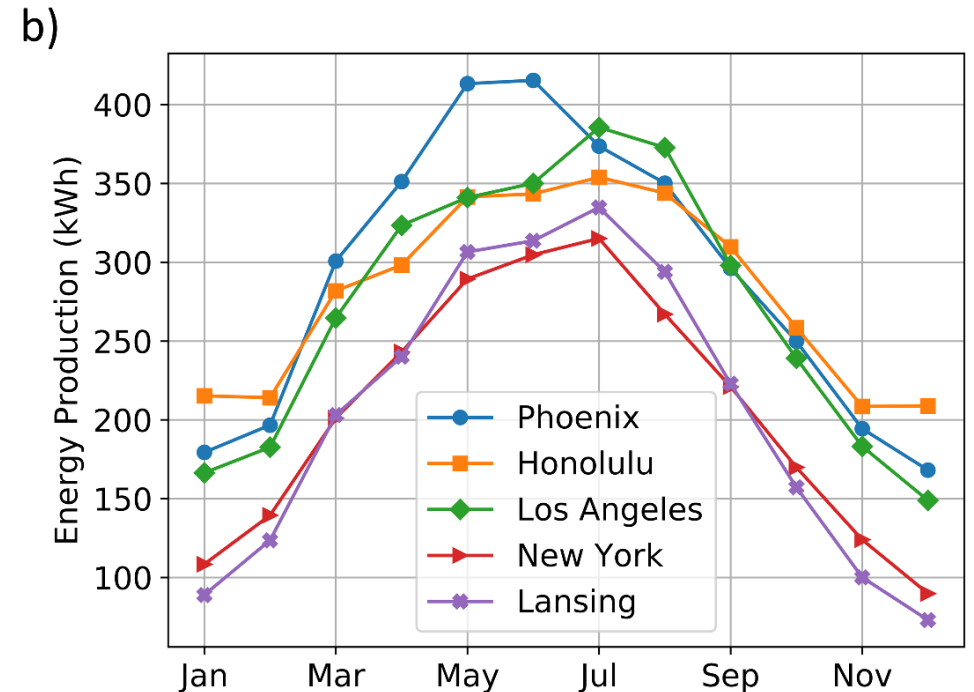
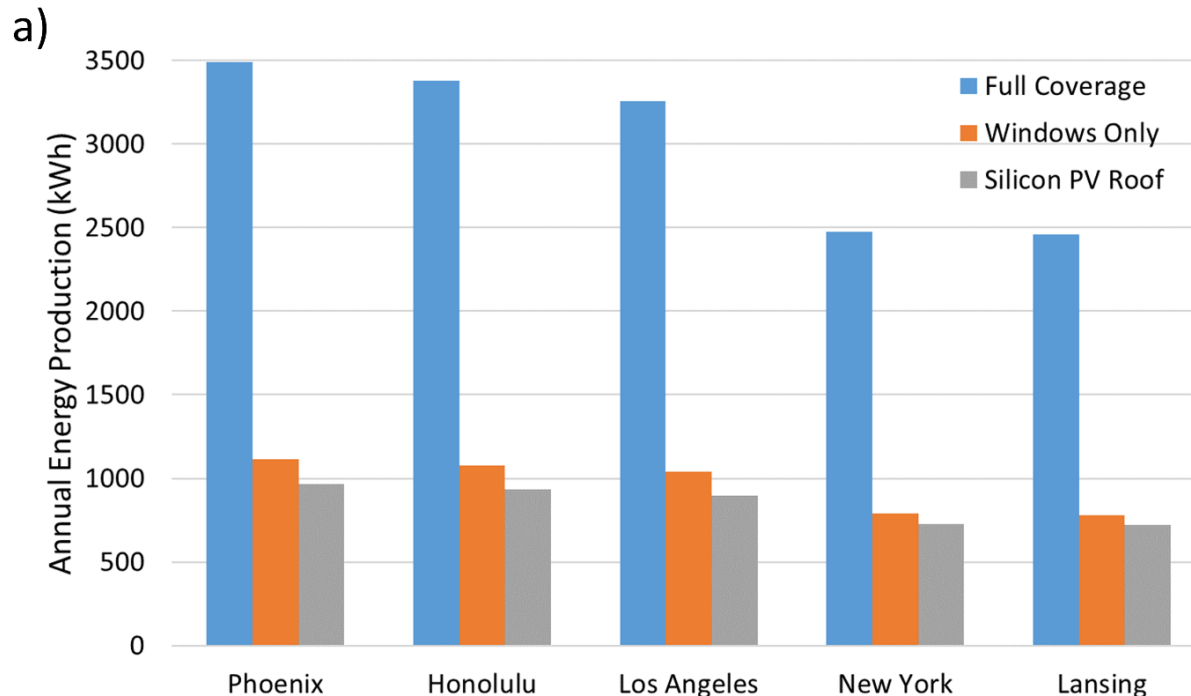
Data from A. Antil et al. Appl. Energy, 2020



Data from Y. Ding et al. Sol. Energy Mater. Sol. Cells, 2015

Potential of Vehicle-Integrated Transparent PVs

- Comparison of energy production for different levels of PV coverage on vehicle and different locations
- Realistic weather and irradiance for each location was taken from TMY data from NSRDB



VIPV Feasibility for HVAC

- Cooling energy needs of an HVAC system during moderate length trips could be entirely supplied by VIPV system

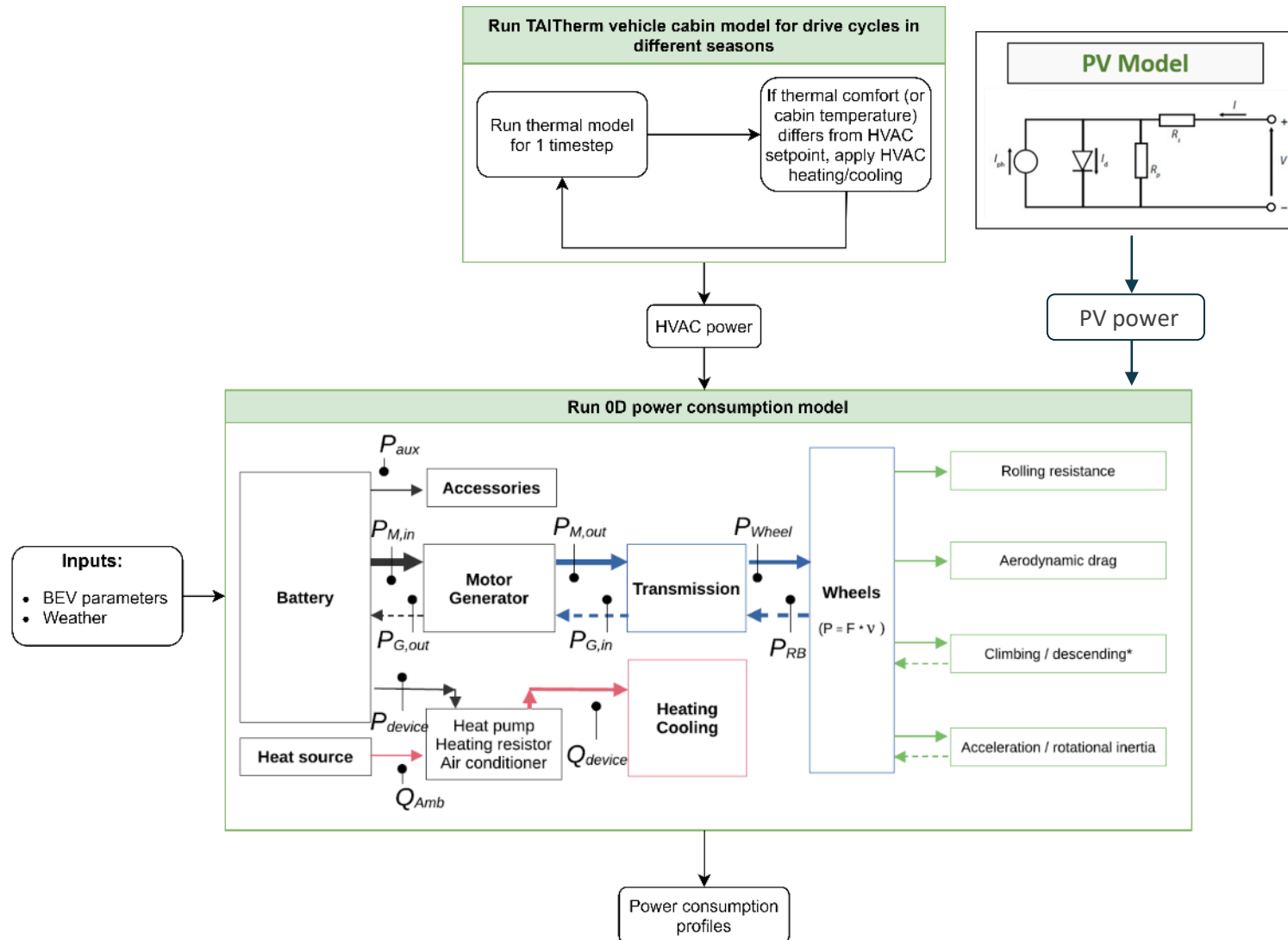
Impact of Desired Vehicle Temperature on EV Energy Consumption		
Outside Temp	Desired Vehicle Temp	Energy Consumption
110 °F	70 °F	1.5-2 kW
110 °F	77 °F	1 kW
110 °F	84 °F	0.5 kW

Data is for an electrically driven heat pump. From <https://avt.inl.gov/sites/default/files/pdf/fsev/auxiliary.pdf>

Modeled Energy Production after 8 hours on a sunny summer day in Phoenix, AZ	
Full TPV Coverage	13.8 kWh
Windows Only TPVs	4.5 kWh
Si PV Roof	3.8 kWh

Example Application: Electric Vehicle Energy Consumption with On-board PVs

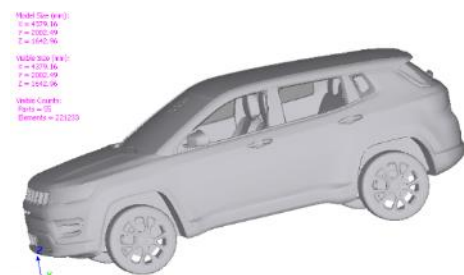
Approach Overview



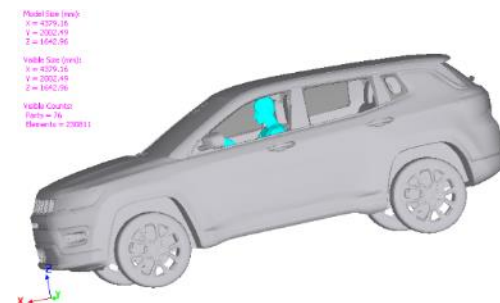
TAITherm Vehicle Modeling Steps



Find garage temperature at start of commute



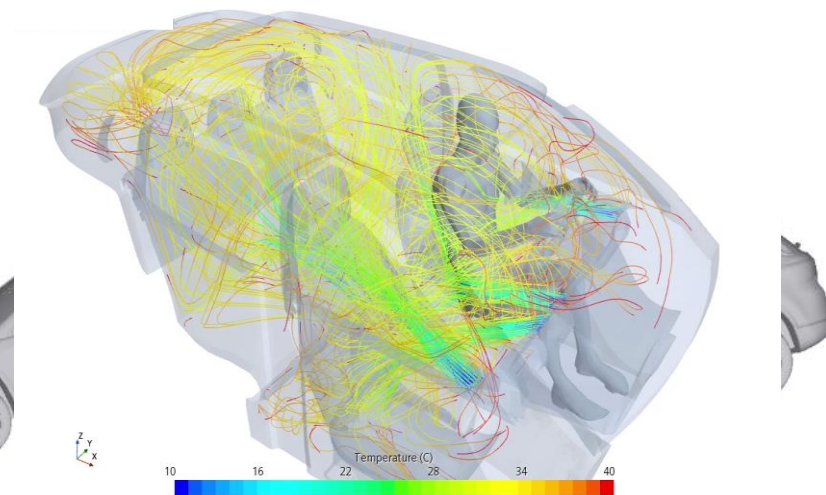
Assume vehicle starts in thermal equilibrium with garage. Run SS on vehicle in bounding box set to garage temperature to set initial conditions.



Transient restart into 40 min. morning commute, adding the driver to the vehicle. HVAC is regulated based on driver's thermal comfort (PMV).



Transient restart into parked car model. Car is parked in an unshaded area for



Transient restart into 40 min. afternoon commute, adding the driver to the vehicle. HVAC is regulated based on driver's thermal comfort.

Library Convection-Flow Speed



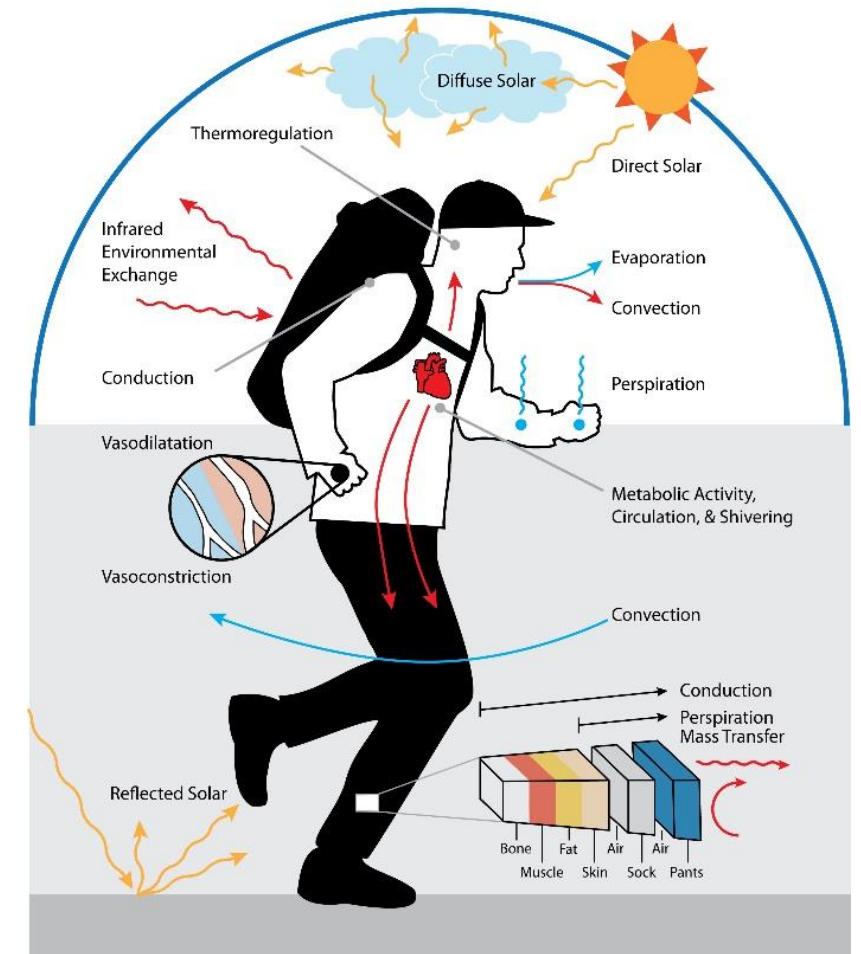
Time-series of HVAC power use and PV power generation

Predicted Mean Vote for HVAC Control

- A comfort metric developed by Fanger (and standardized by ISO7730)
- Predicts mean value of votes of large group of persons on a 7-point thermal sensation scale ranging from -3 (cold) to +3 (hot)
- Considers activity level, clothing resistance, air temperature, mean radiant temperature, air velocity, and relative humidity
- An output of the TAITherm Human Thermal Model

TAITherm Human Thermal Model

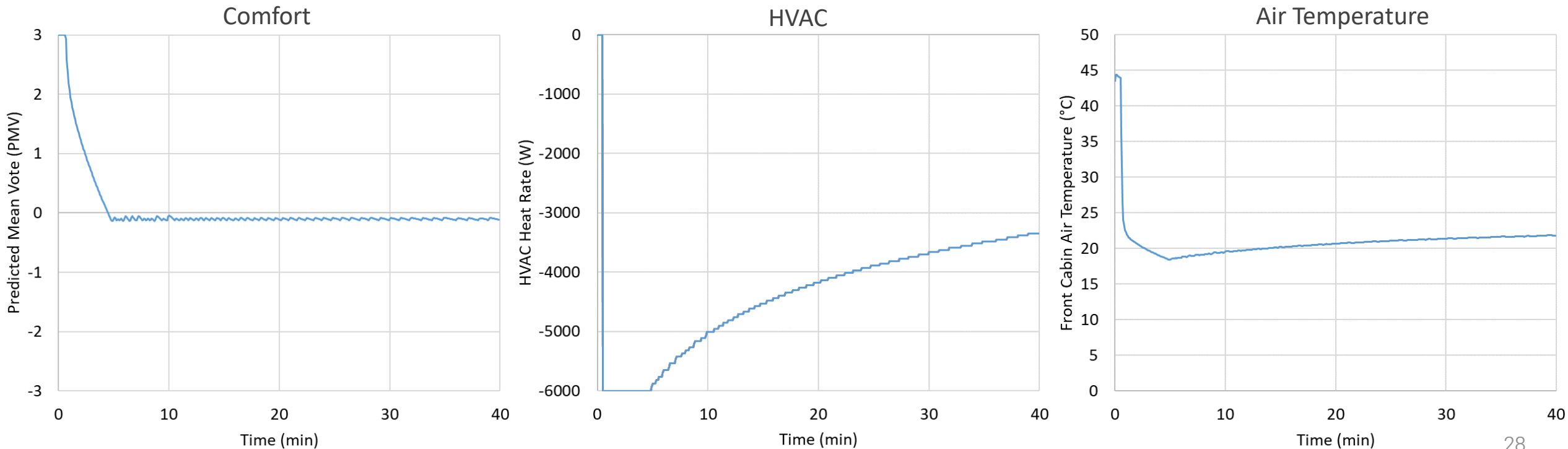
- 20 body segments, with layers representing tissues
- Solves bio-heat transfer equation
- Predicts tissue, blood, and core temperatures under:
 - Varying environmental conditions
 - Varying activity levels
 - Adjustable clothing levels



Mechanisms considered by TAITherm Human Thermal Model

Human Comfort for HVAC Power Consumption

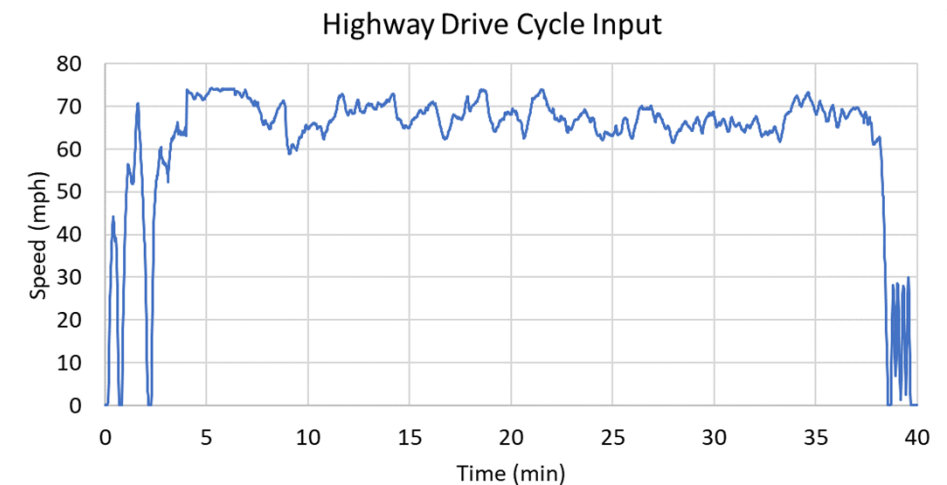
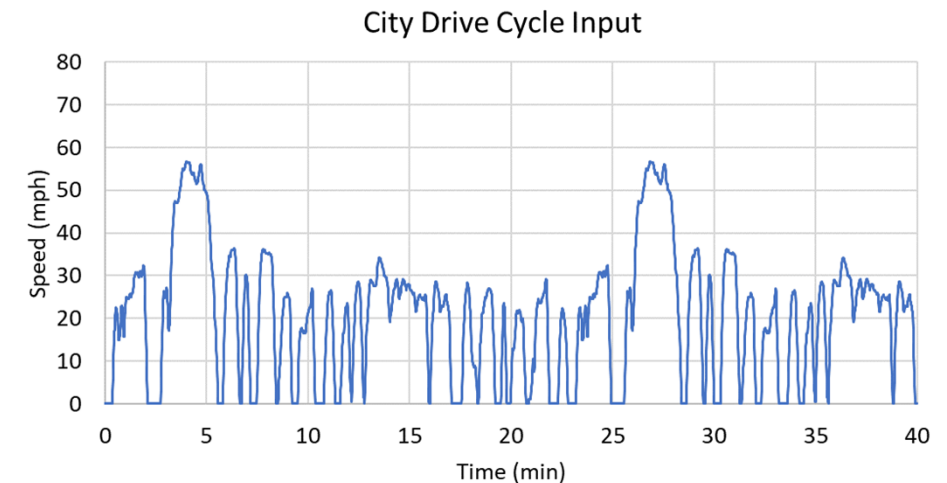
- HVAC control algorithm implemented in a User Routine
- $Q_{HVAC} = f(PMV, d(PMV)/dt)$, $v_{air} = f(PMV, d(PMV)/dt)$



EV Power Model Inputs

- EV parameters
- Drive cycle (speed time-series)
- Weather
- HVAC load and PV power from TAlTherm cabin model

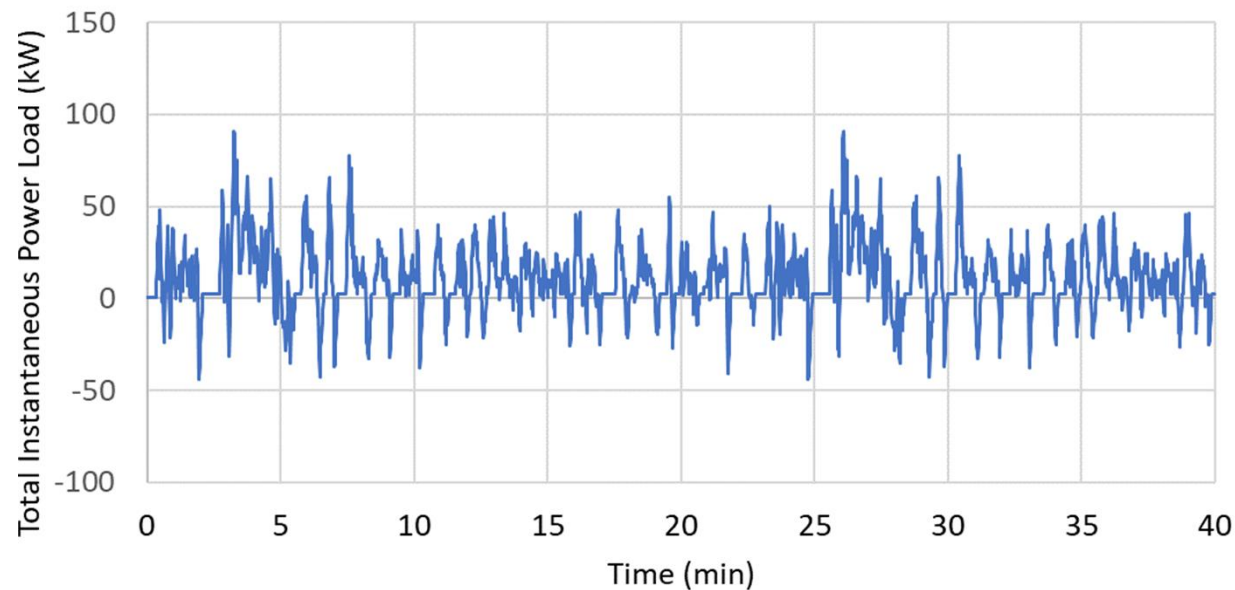
Parameter	Value
Curb Weight	3000 kg
Gear Ratio	8.0
Frontal Area	2.88 m ²
Drag Coefficient	0.371
Max Power	220 kW (300 hp)
Battery Discharging Efficiency	95%
Battery Charging Efficiency	90%
Transmission Efficiency	95%
HVAC Coefficient of Performance	2
Auxiliary Power	0.3 kW



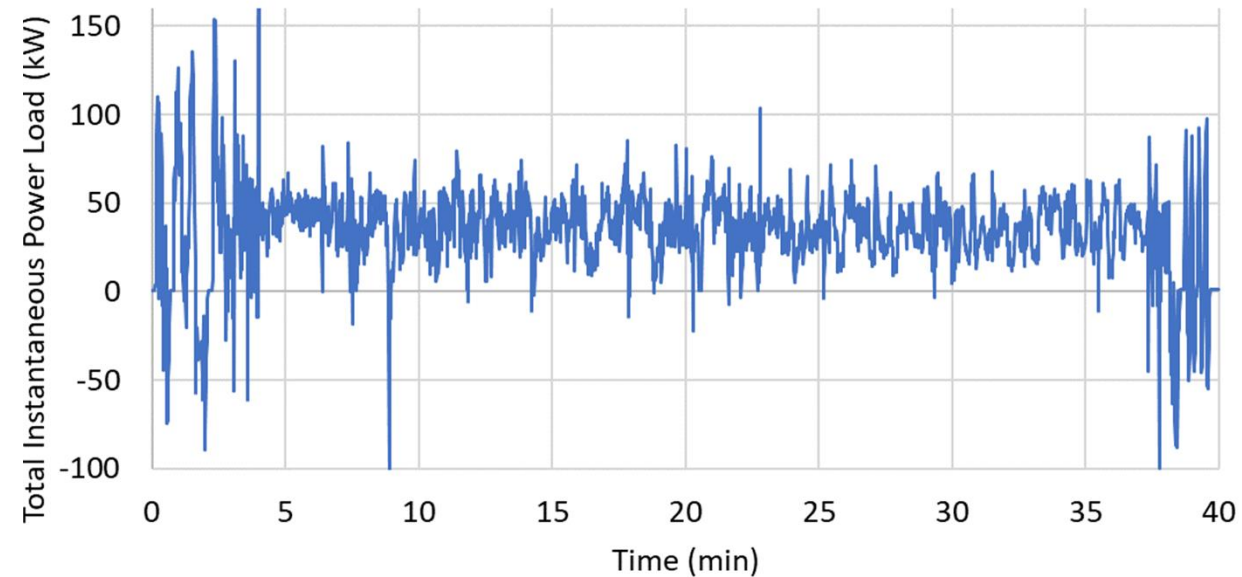
EV Power Model Outputs

- Total power load time-series
- Power loads and losses from individual components of EV

City Drive Cycle Calculated Power Load

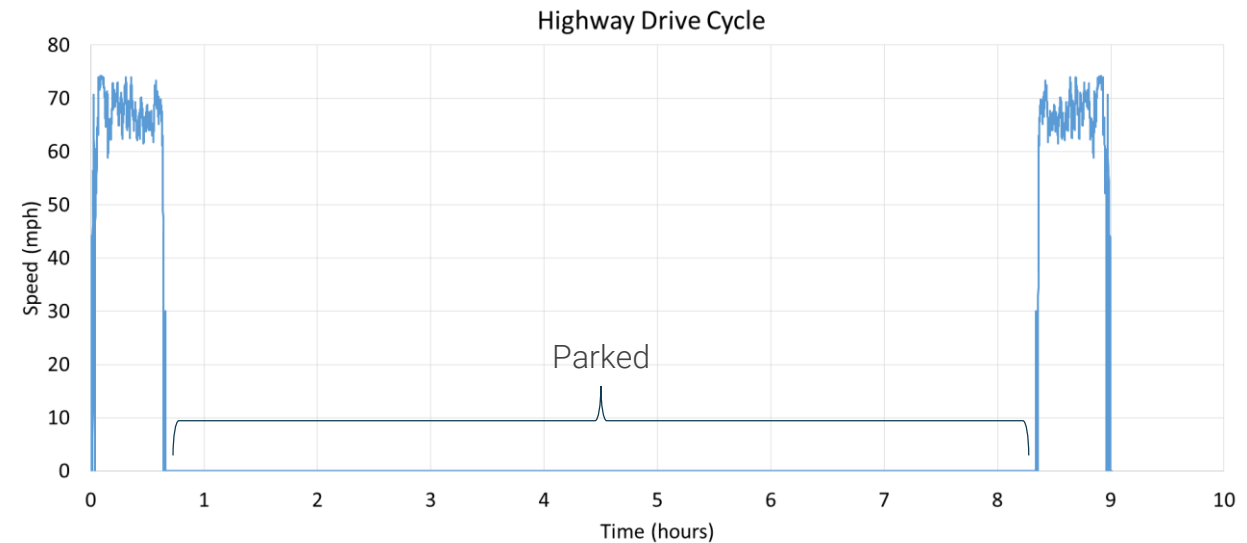
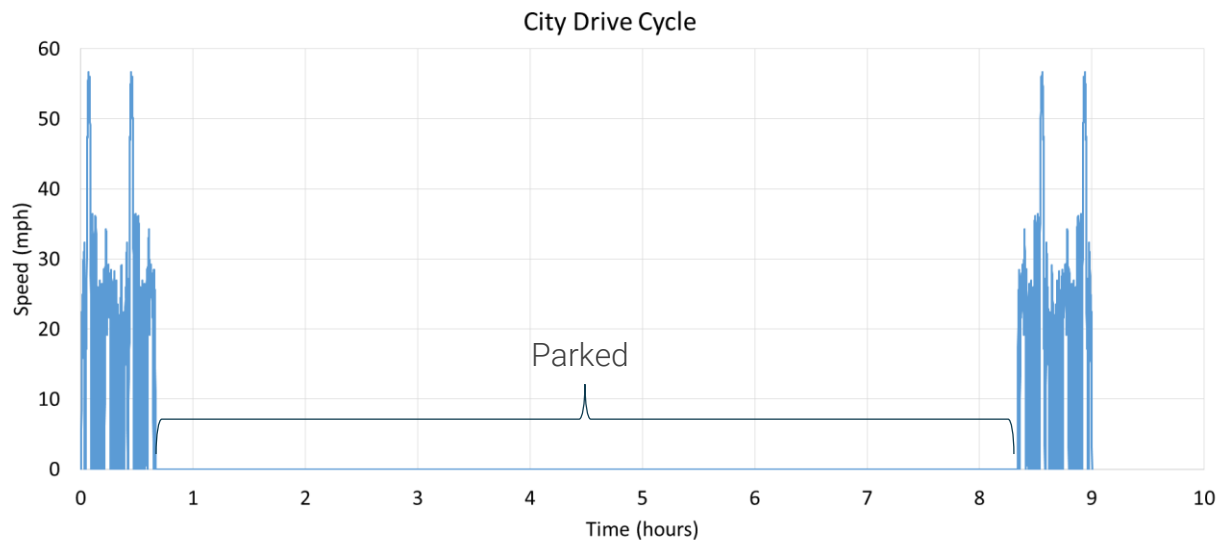


Highway Drive Cycle Calculated Power Load



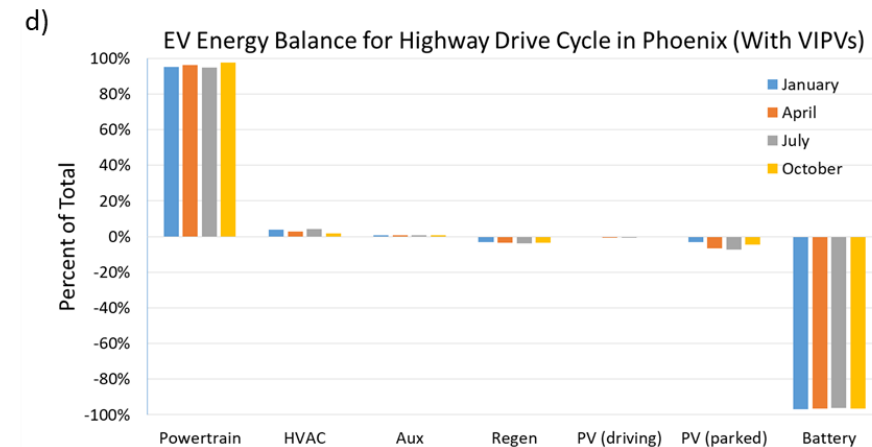
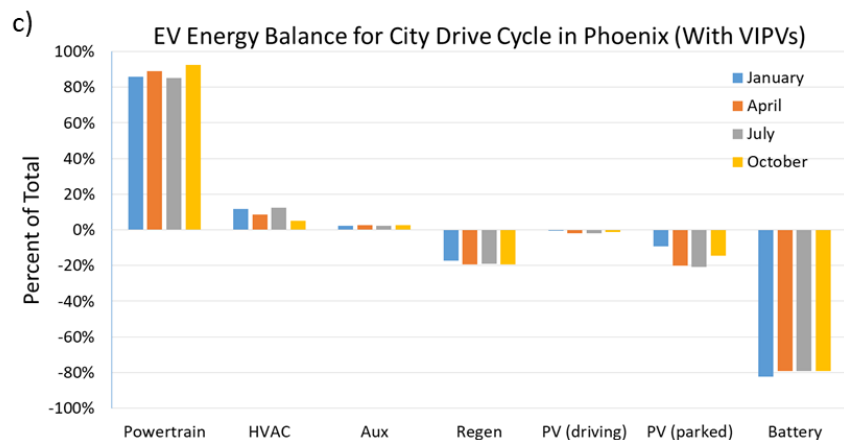
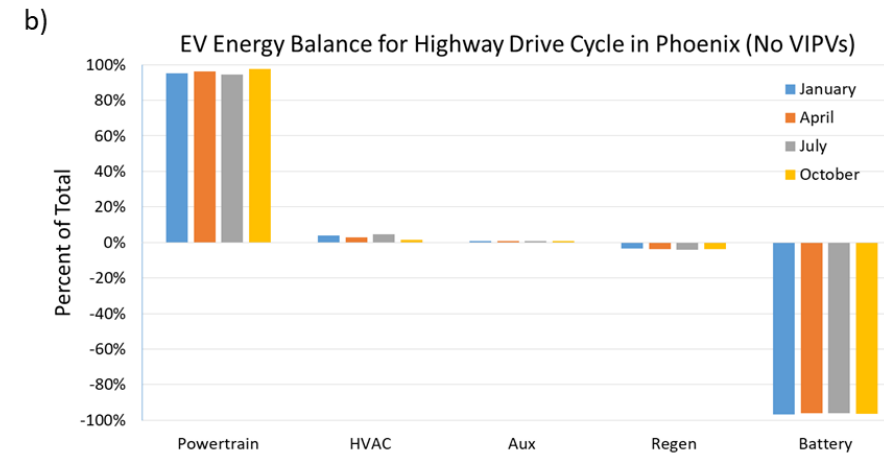
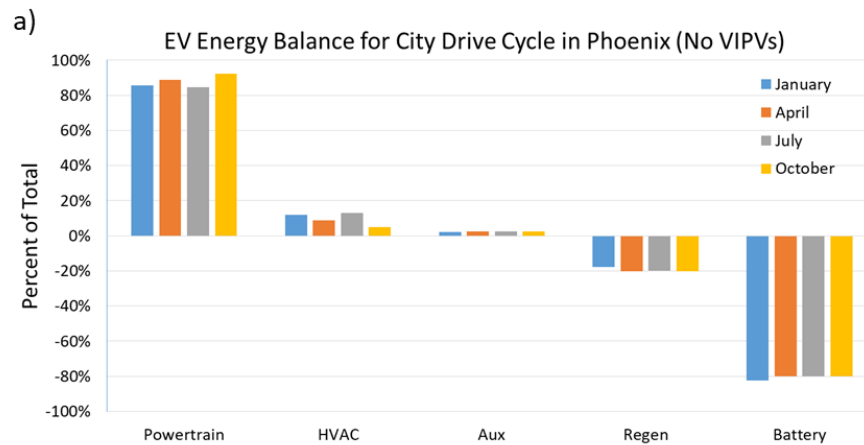
Case Study

- Modeled EV energy balance over 1 day for 2 drive cycles in each of 4 seasons in Phoenix, Arizona and Detroit, Michigan
 - 40 min morning commute + 8 hours parked outside + 40 min afternoon commute



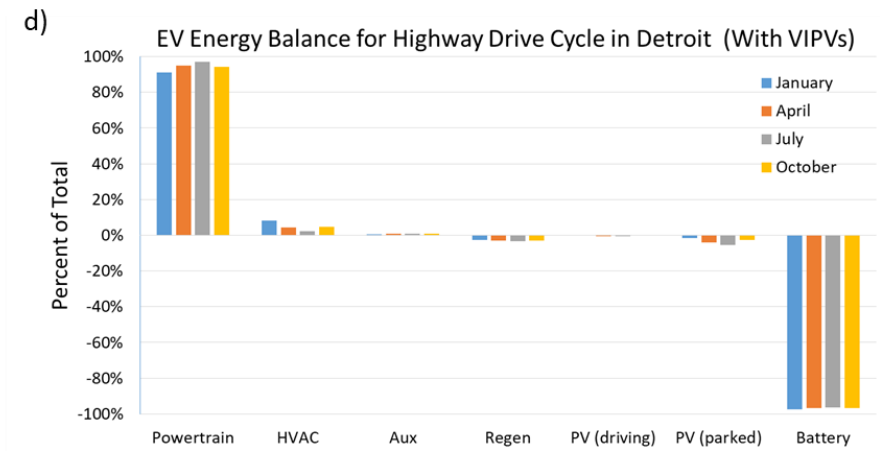
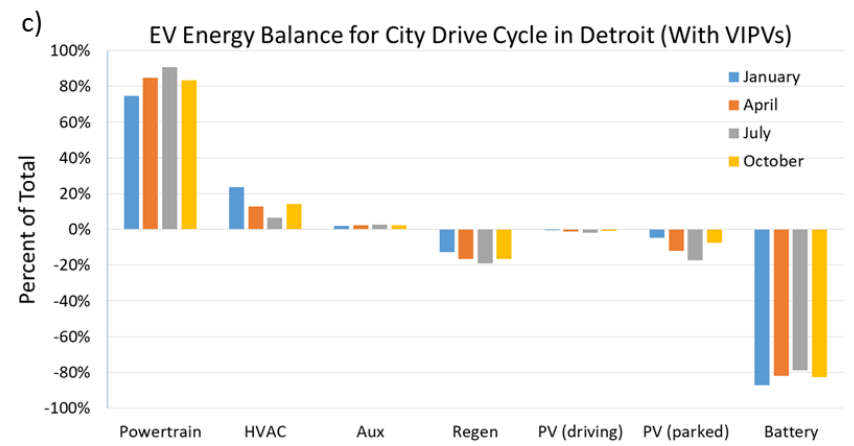
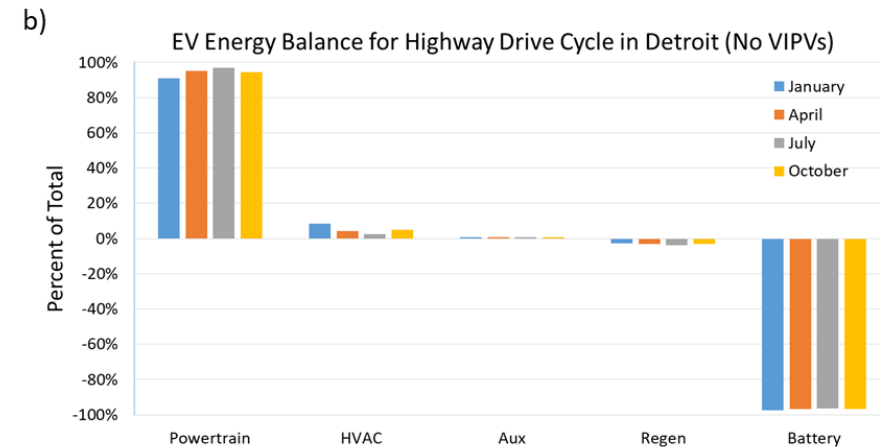
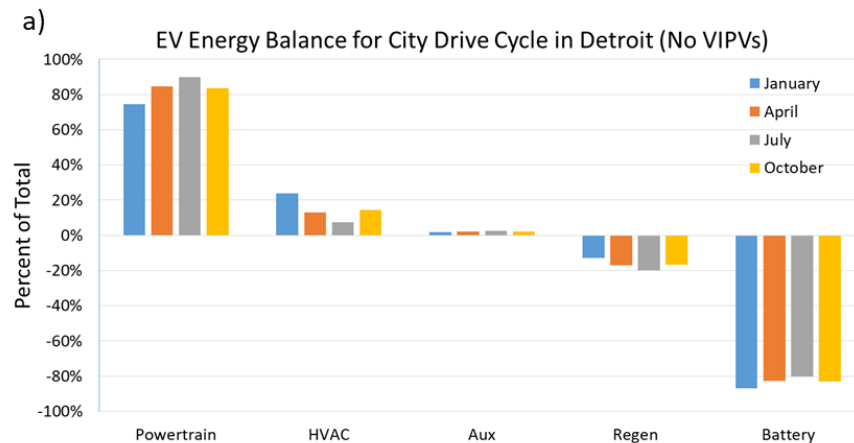
Results: Phoenix, Arizona

- VIPVs generate 10-23% and 3-8% of consumption (or 4-11 km of range) of city and highway drive cycles, respectively



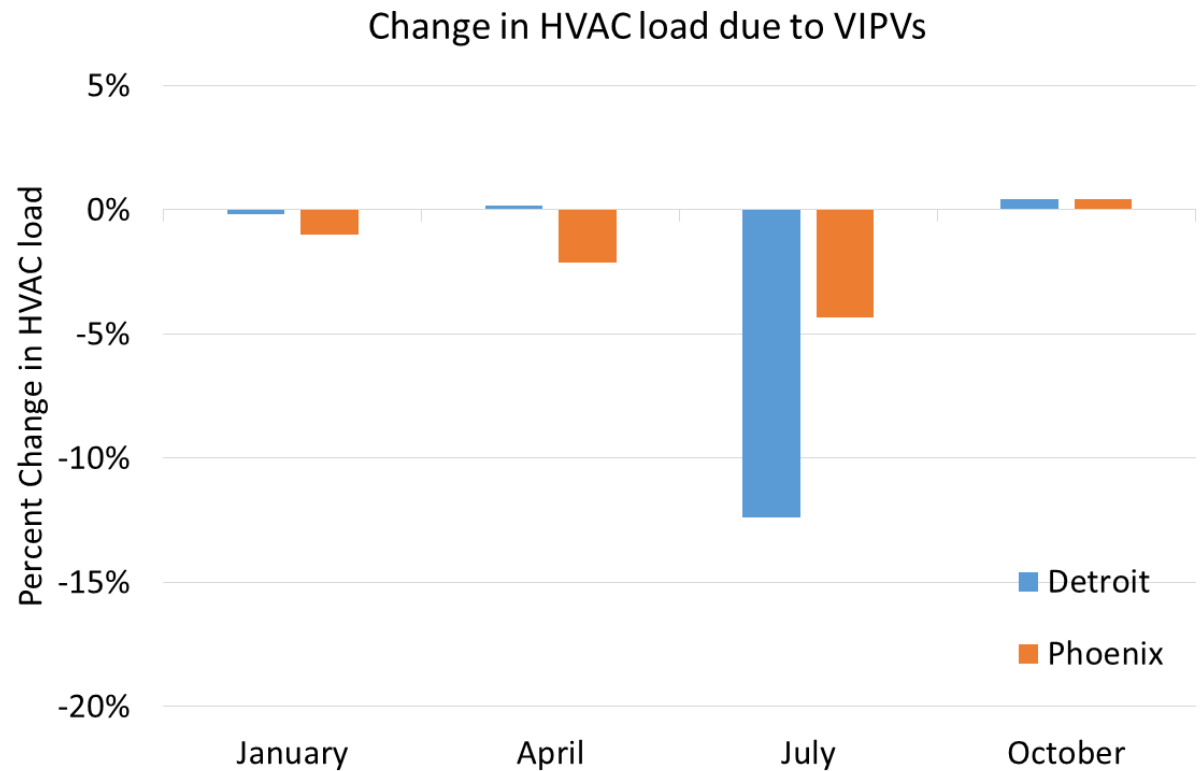
Results: Detroit, Michigan

- VIPVs generate 5-19% and 2-7% of consumption (or 2-9 km of range) of city and highway drive cycles, respectively



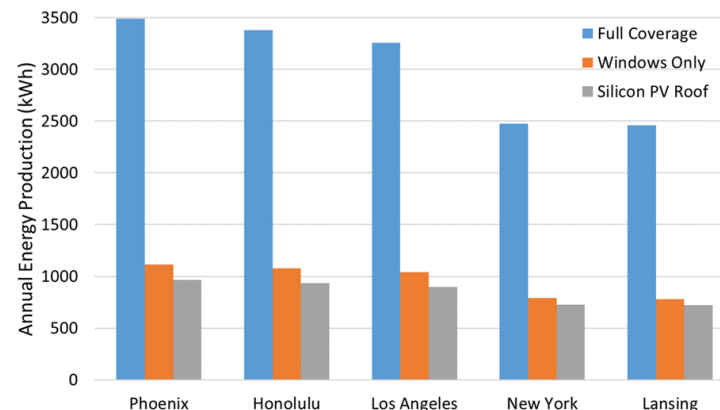
Results: Impact on HVAC Load

- HVAC load significantly reduced in cooldown scenarios due to VIPV's converting some of solar radiation to electricity instead of heat



Conclusions

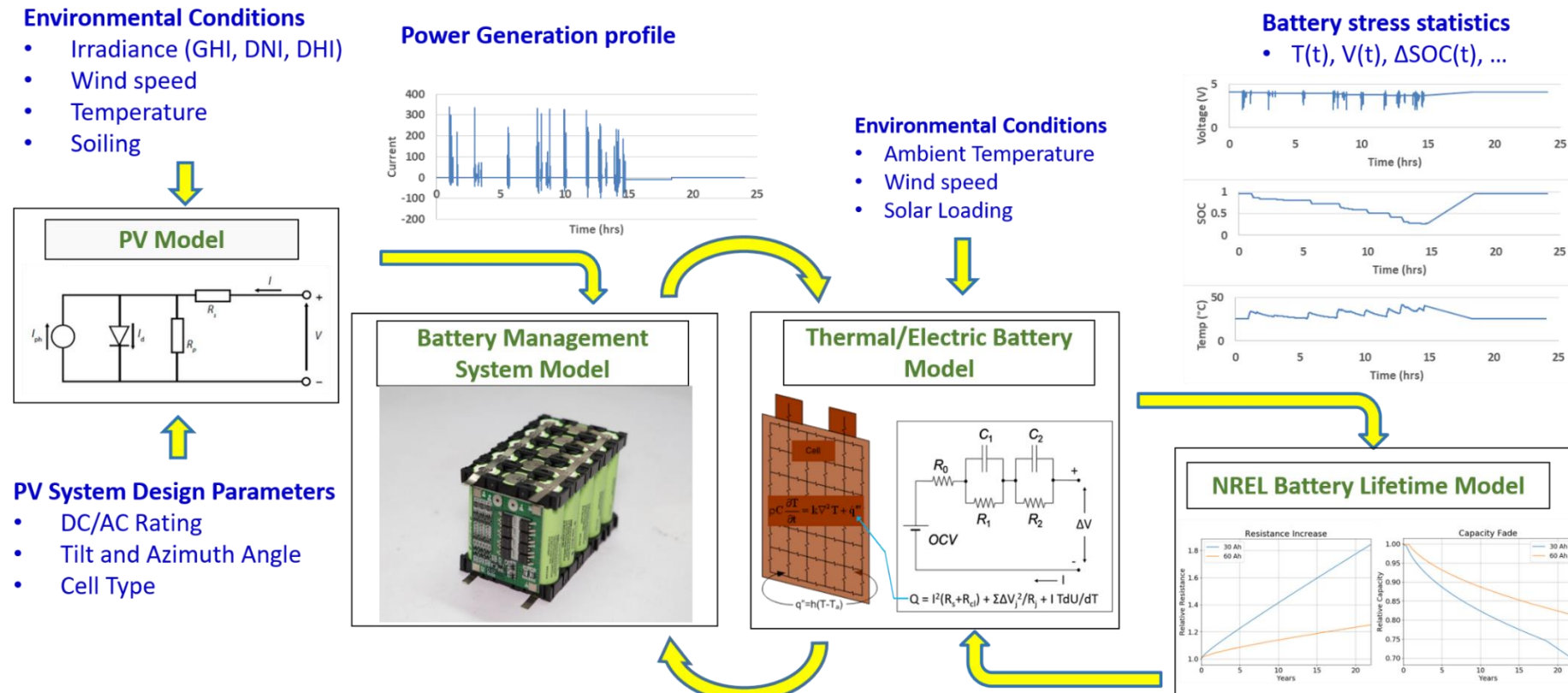
- We demonstrated how TAItherm could be used for EV energy consumption with consideration of VIPVs and HVAC loads
- For a full size SUV, with non-optimized aerodynamics (0.371 drag coeff.), amount of range extension expected from integration of conventional solar cells into the roof and hood is 2-11 km per day
 - More aerodynamic or lighter weight vehicles would get more range from VIPVs
 - ~3x more energy could be gained by achieving full vehicle coverage



Future Work

- PV thermal-electrical model and EV energy model could be coupled with thermal-electrical battery performance and lifetime models for more detailed analysis.

Potential Future Modeling Approach



References

1. <https://www.seia.org/solar-industry-research-data>
2. Precedence Research, “Solar Photovoltaic Market Size to Worth Around US\$250.6 BN by 2030”, 2022.
3. T. Golubev, “Multi-physics modeling and simulation of photovoltaic devices and systems”, PhD thesis, Michigan State University, 2020, <https://d.lib.msu.edu/etd/49526>
4. T. Golubev and R. R. Lunt, “Evaluating the Electricity Production of Electric Vehicle-Integrated Photovoltaics via a Coupled Modeling Approach”, 2021 IEEE 48th Photovoltaic Specialists Conference
5. T. Golubev et al., “Analyzing the Impact of On-Board Photovoltaics on Electric Vehicle Energy Consumption”, Accepted to 2022 IEEE Transportation Electrification Conference, Anaheim, CA, June 2022.
6. P. O. Fanger, “Thermal Comfort”, Danish Technical Press, Copenhagen, 1970.



Thank you

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