

Economic Analysis and Optimal Sizing of Battery Storage for Residential Consumers with Solar

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Key Question

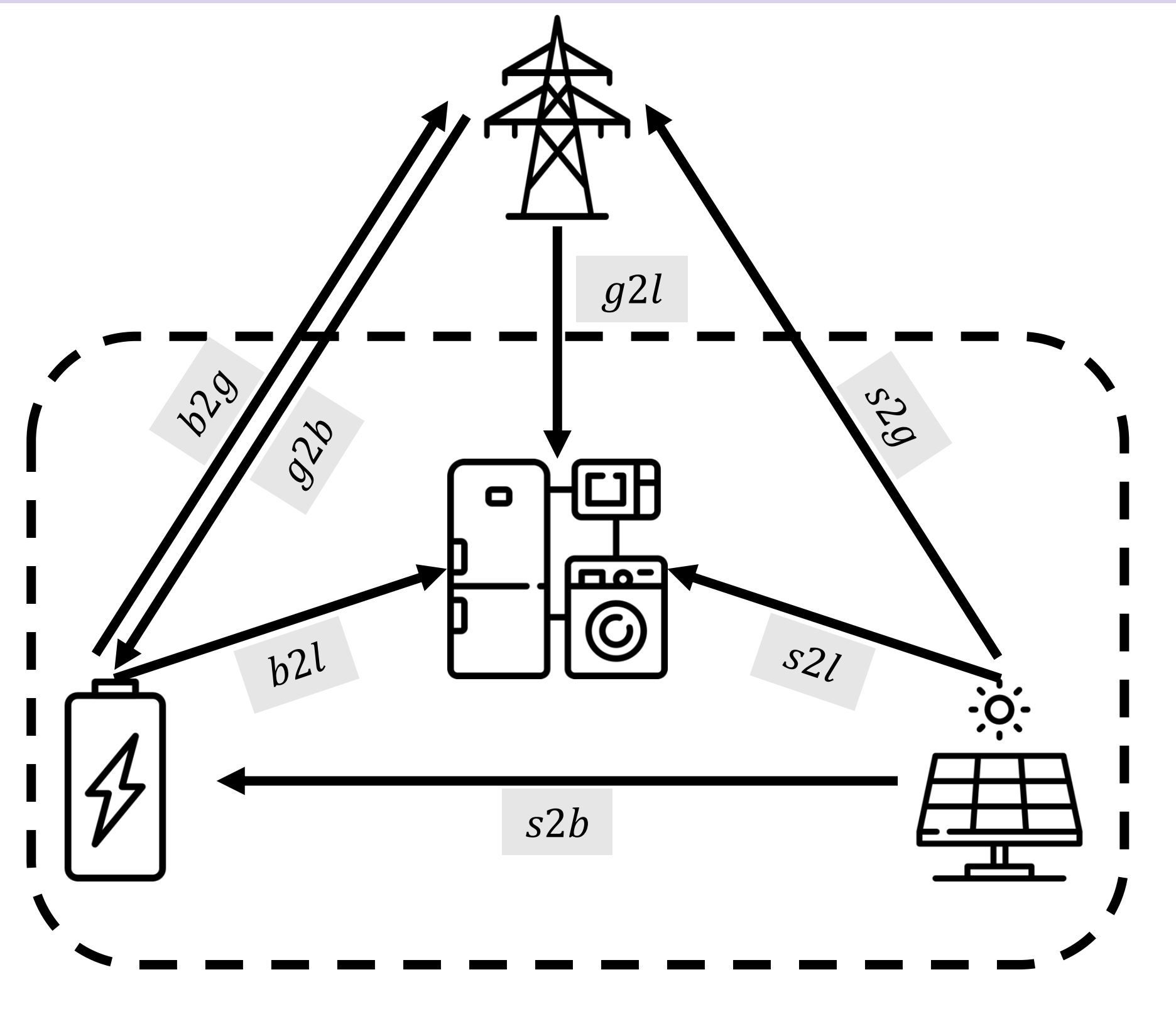
Is battery storage economically viable for residential solar owners?

Our Answer: YES! With proper sizing, returns of 10-16% IRR are achievable.

What We Did

- **Developed** a comprehensive framework combining optimal control with economic analysis
- **Optimized** battery operations hourly using forecast solar generation
- **Evaluated** lifetime economics using Internal Rate of Return (IRR)
- **Identified** optimal battery specifications for maximum returns on investment.

System Overview



Key Components & Variables:

- **Solar PV:** Generation S_t at time t (weather-dependent)
- **Battery:** Energy capacity E_{cap} , Power capacity P_{cap}
 - State of Charge: SOC_t
 - Efficiency: η_c (charging), η_d (discharging)
- **Grid:** Buying price π_t^{buy} , Selling price π_t^{sell}
- **Load:** Household consumption L_t at time t

Decision Variable: Battery charge/discharge power P_t^{batt} at each hour t (negative = charging, positive = discharging)

Why Battery Degradation Matters

The Challenge: Complex real-world battery aging model.

Key Degradation Factors:

- **Cycle degradation:** Deeper discharge cycles = faster wear
 - Non-linear relationship: doubling depth $\approx 4\times$ degradation
 - Tracked using Rainflow counting algorithm (no closed-form solution!)
- **Calendar aging:** Time-based degradation ($\sim 2\%$ /year)
- **Early-life effects:** Higher initial degradation (SEI formation)

Result: Battery reaches end-of-life (70% capacity) after 8-12 years depending on usage

How We Optimize: Stochastic MPC

The Approach: Look ahead 24 hours, act on the current hour.

At Each Hour:

- 1 Generate multiple solar forecast scenarios for next 24 hours
- 2 Solve for cost optimization considering all scenarios and operating constraints
- 3 Execute only the first hour's decision
- 4 Move on to the next hour with updated forecasts

Optimization Objective:

$$\min \mathbb{E} \left[\sum_t \left(\pi_t^{buy} P_t^{buy} - \pi_t^{sell} P_t^{sell} + \alpha_{deg} |P_t^{batt}| \right) \right]$$

Total Cost = Electricity Bought - Solar Sold + Degradation Cost

Subject to:

- Power balance between solar PV, battery, load and grid transactions.
- Battery operation within physical limits.
- Approx. linear degradation cost coefficient: α_{deg} (tunable parameter)

Solar Forecast: Managing Uncertainty

The Challenge: Generate solar forecasts based on historical data: **mean-reverting** process with parameter fitting to preserve temporal correlations.

Our Solution: Ornstein-Uhlenbeck (OU) process for forecast errors

Forecast Generation:

$$\epsilon_t = (1 - \phi) \cdot \epsilon_{t-1} + \sigma \cdot \xi_t$$

$$\log(F_t) = \log(H_t) + \epsilon_t$$

where H_t = historical solar, F_t = forecast, $\xi_t \sim \mathcal{N}(0, 1)$

- **Mean reversion parameter** $\phi = 0.37$: Controls error persistence
- **Volatility** $\sigma = 0.42$: Controls forecast uncertainty
- Generates 10 realistic scenarios in each hour's stochastic MPC problem

Economic Evaluation Methodology

From Optimization to Investment Analysis:

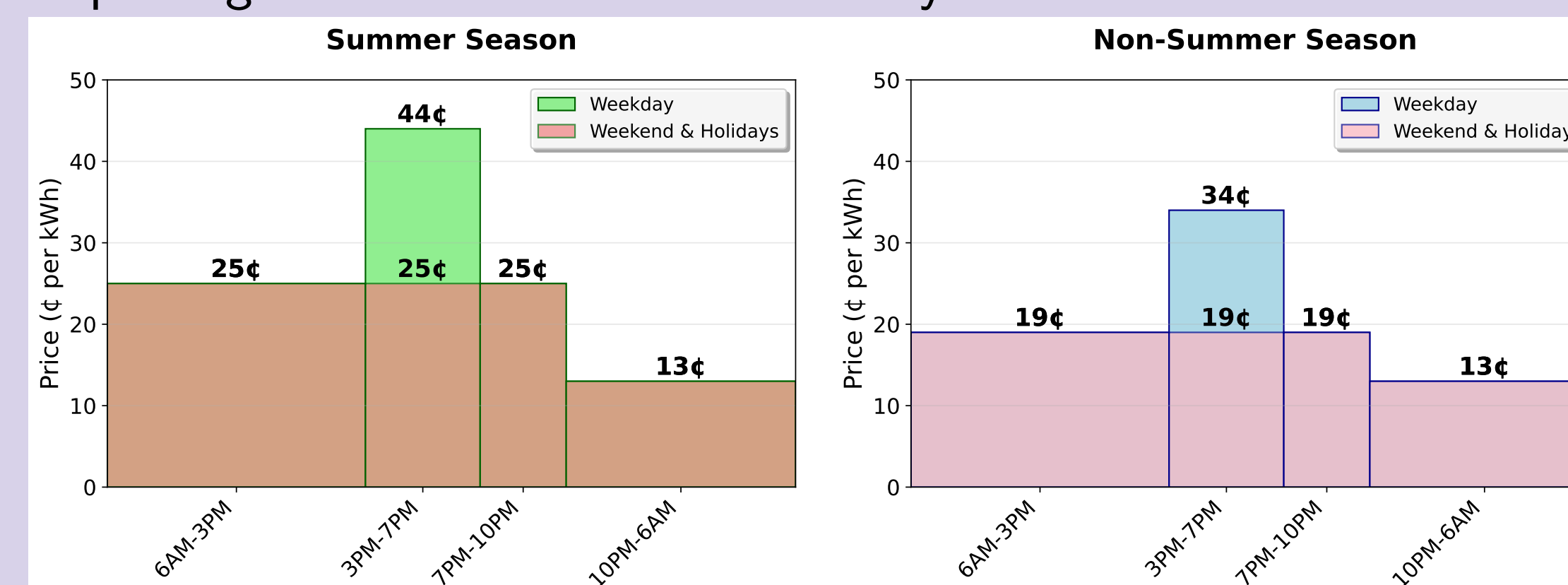
- 1 **Year 1 Simulation:** Full year of optimized operations
 - Net revenue: $R_{annual} = R_{with\ battery} - R_{baseline}$
 - Degradation: ΔD_{annual} via Rainflow algorithm
- 2 **Multi-year Projection:** Account for capacity fade
 - Year y capacity: $C(y)$ based on degradation
 - Year y revenue: $R_{annual} \cdot C(y)$

End-of-life when capacity drops below 70%

Real-World Case Study: Austin, Texas

Dataset: One year of real-world hourly data from Pecan Street

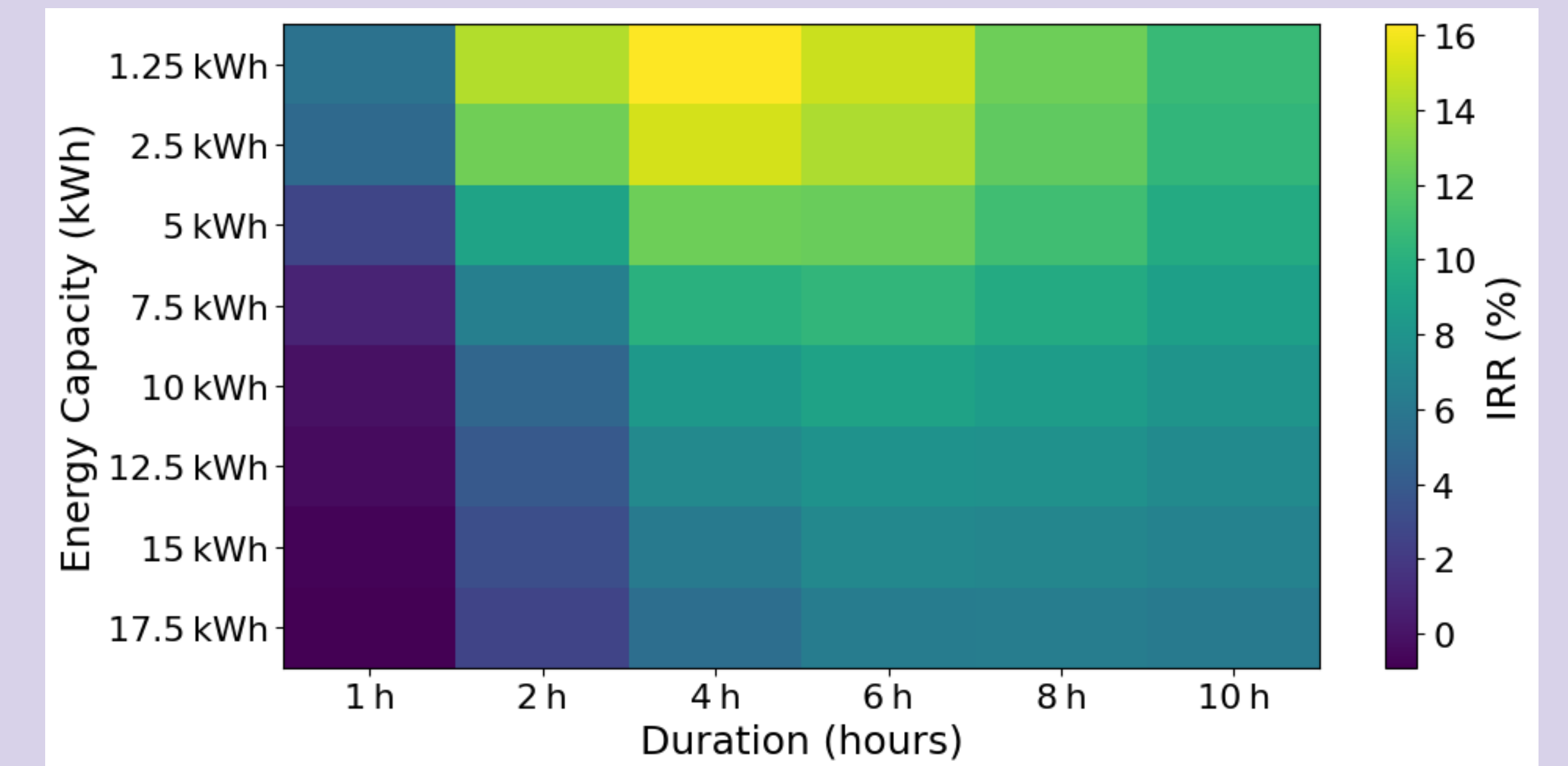
- Residential consumption scaled to US average of 30 KWh per day.
- Rooftop PV generation scaled to 7.5 KW system.



- **Peak hours (3-7 PM):** Up to 44 cents/kWh summer
- **Off-peak:** As low as 13 cents/kWh
- **Solar selling price:** 9.91 cents/kWh (flat rate, no arbitrage opportunity)
- **Battery costs by 2040:** \$200/kWh (energy storage) + \$300/kW (inverter)
- **Opportunity:** Store cheap solar/off-peak energy for expensive peak hours!

Main Results: Optimal Battery Sizing

Sweet Spot: 2.5-7.5 kWh capacity, 4-6 hour duration
Returns: 10-16% IRR (better than many investments!)



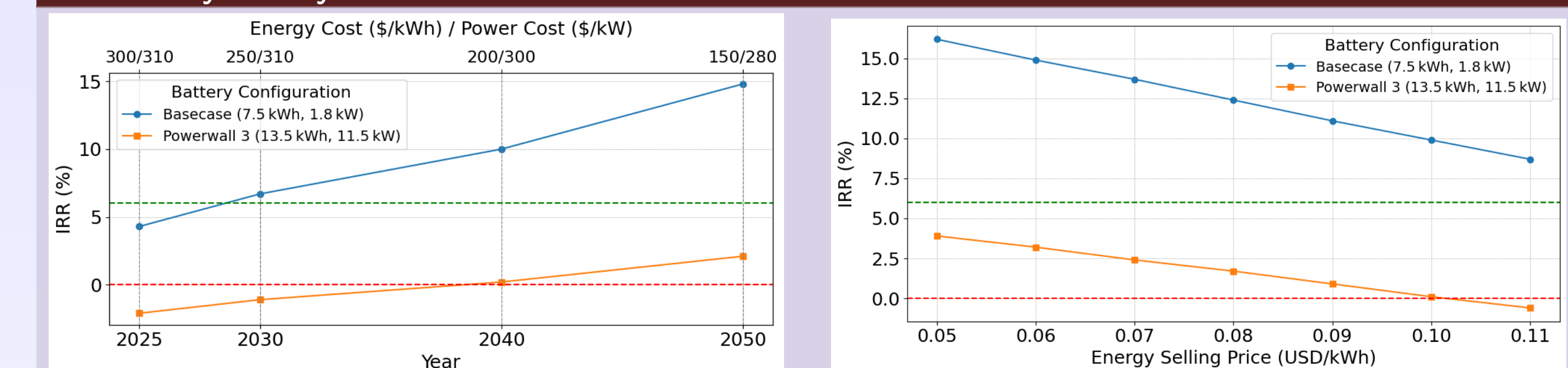
Key Investment Benchmarks:

- **Small battery (2.5 kWh, 4h):** 15.2% IRR with \$680 investment cost
- **Medium battery (7.5 kWh, 4h):** 10.0% IRR with \$2,040 investment cost
- **Powerwall 3 specs (13.5 KWh, 11.5 KW):** 0.2% IRR with \$6,150 investment cost. Oversized for average homes!

Why smaller is better:

- Fully utilized daily (no wasted capacity)
- Lower initial investment
- Diminishing returns with size

Sensitivity Analysis



Future Cost Projections

Key Insights:

- Battery costs dropping \rightarrow IRR improving
- Lower solar selling prices \rightarrow Higher battery value
- By 2030: Even large batteries become viable ($>6\%$ IRR)

Take-Home Messages

For Homeowners:

- Battery storage profitable with proper sizing
- Don't oversize - smaller batteries give better returns
- 4-6 hour duration is optimal for most homes

For Researchers:

- Accurate degradation modeling crucial for economics
- Stochastic MPC effectively handles solar uncertainty
- IRR provides fair comparison across configurations

Acknowledgments

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