

# Enhanced Resource Adequacy Representations for Power System Capacity Expansion Planning

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Gord Stephen

March 29, 2023

Doctoral Candidate

Department of Electrical and Computer Engineering

University of Washington

# Capacity Expansion and Resource Adequacy

## Capacity Expansion Planning

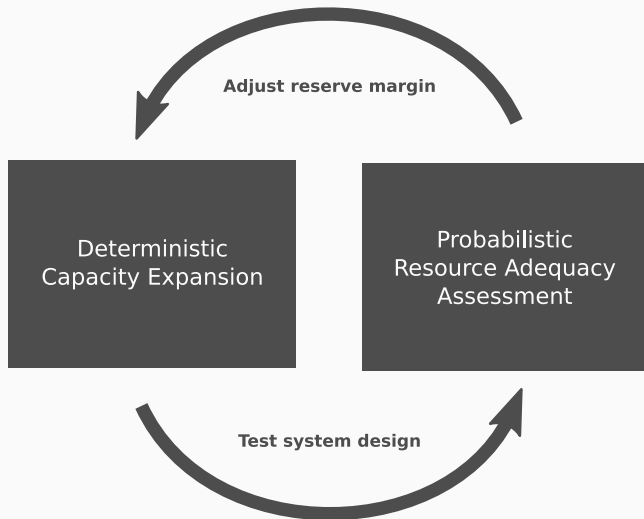
- Longer-term planning optimization
- Computationally demanding, even in the deterministic case
- Usually only considers limited number of representative operating periods
- Resource adequacy is a key constraint

## Resource Adequacy Assessment

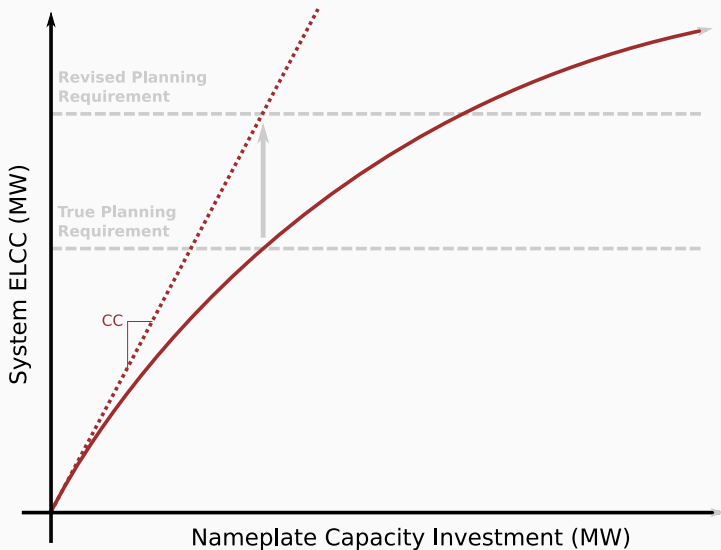
- Shorter-term operations simulation
- Preferably probabilistic, considering many alternative realizations of future system state
- Every hours matters

How do we reconcile these paradigms to plan economically-efficient, resource adequate systems without sacrificing computational tractability?

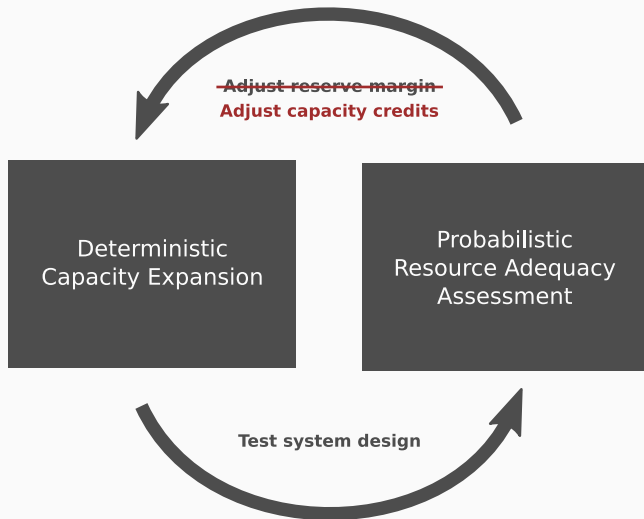
# Planning Reserve Margin Iteration



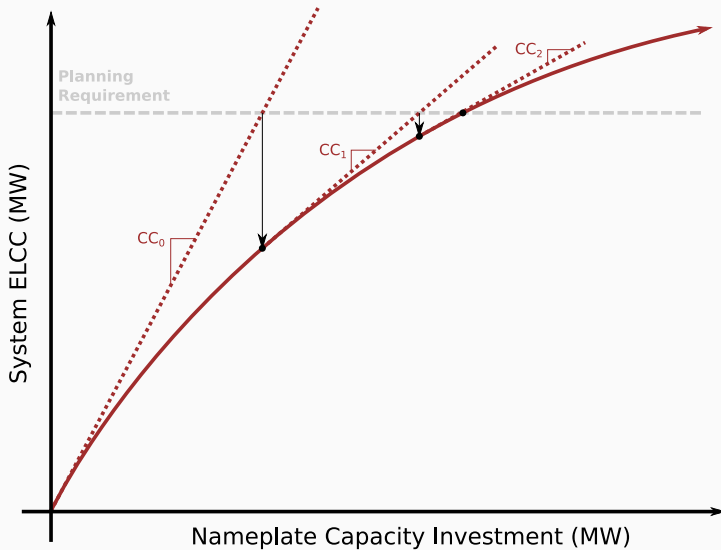
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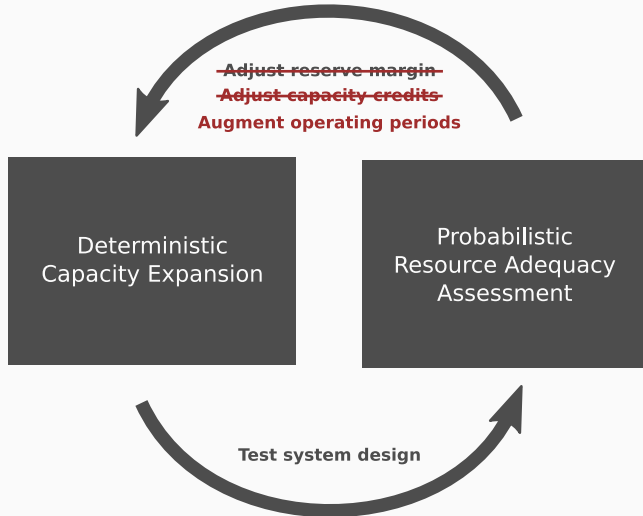
# Capacity Credit Iteration



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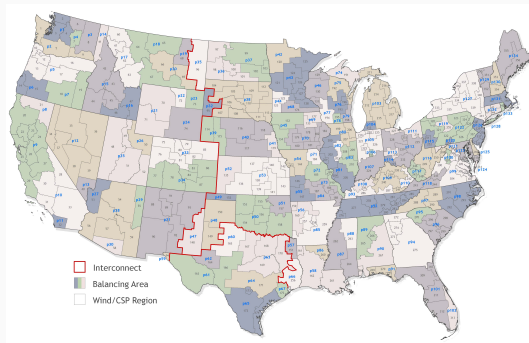


# Risk Period Iteration



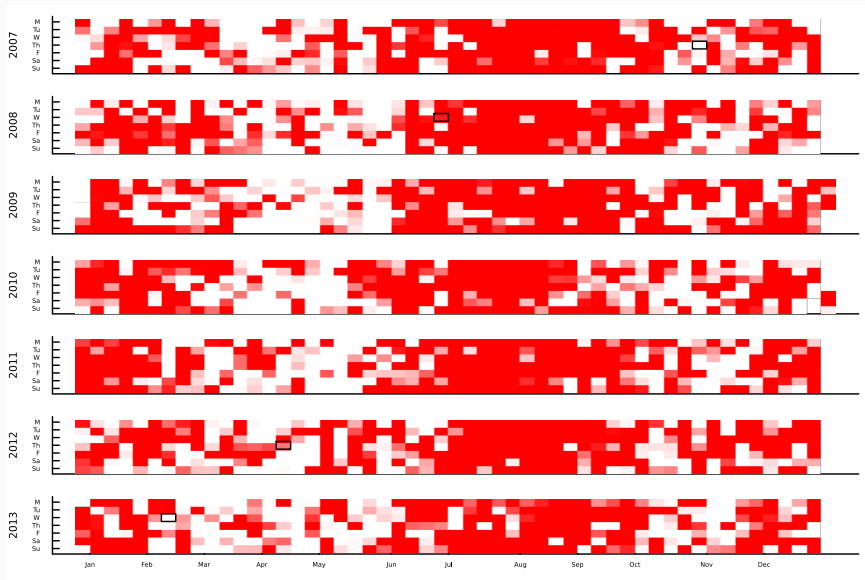
# Risk Period Iteration: example application

- Adequacy stress test: plan a greenfield, 100% wind/solar/storage system
- Seven year resource+load dataset with transmission constraints between 134 regions of contiguous U.S. (derived from ReEDS inputs)

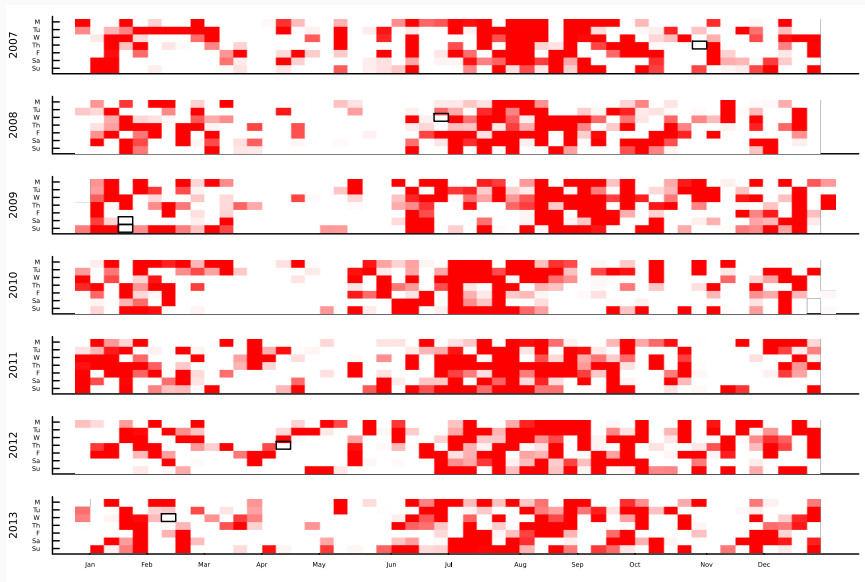




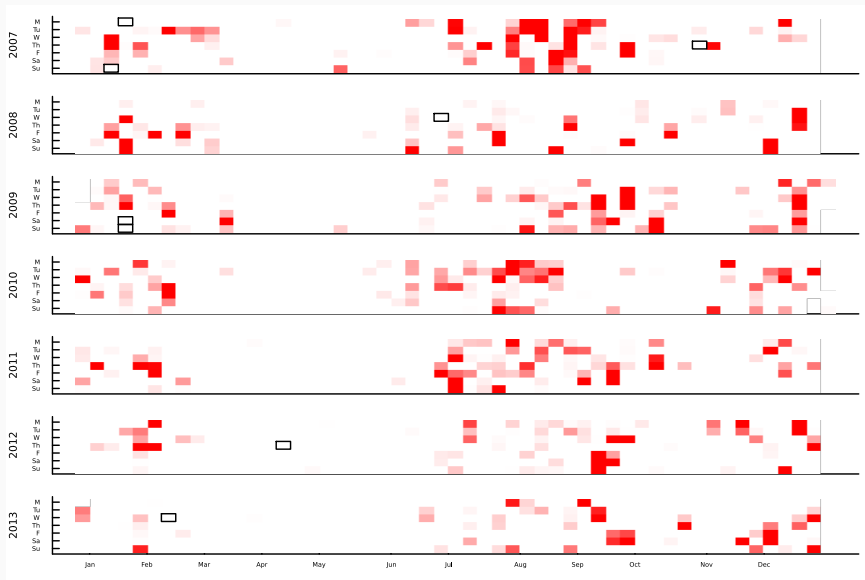
# Risk Period Iteration: discovering days to drive adequacy investment



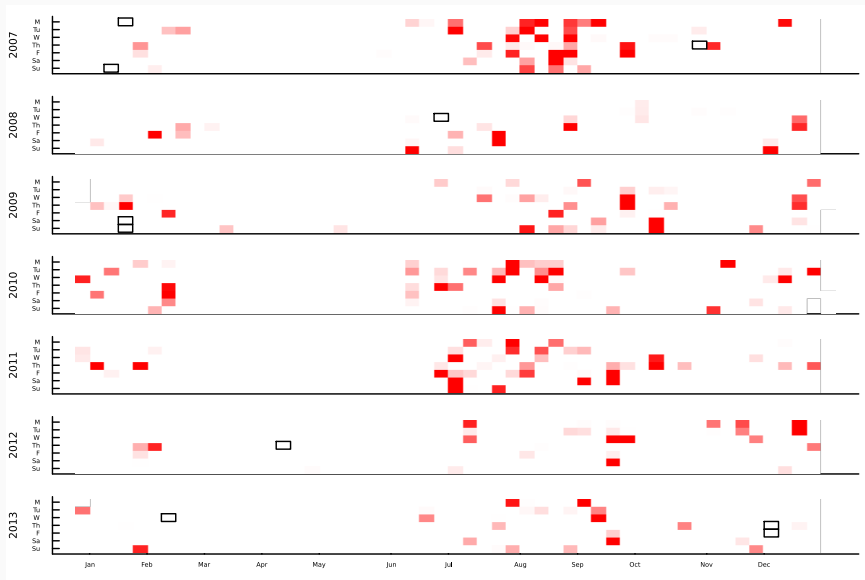
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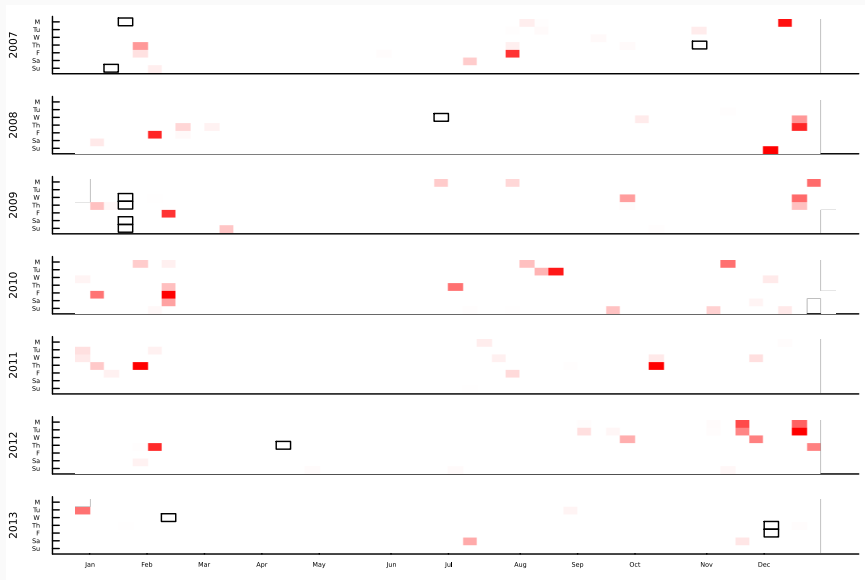
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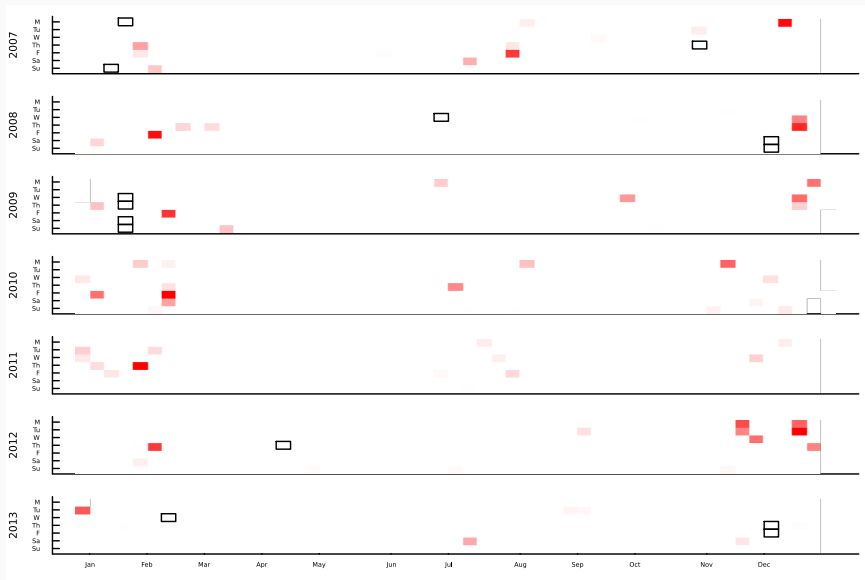
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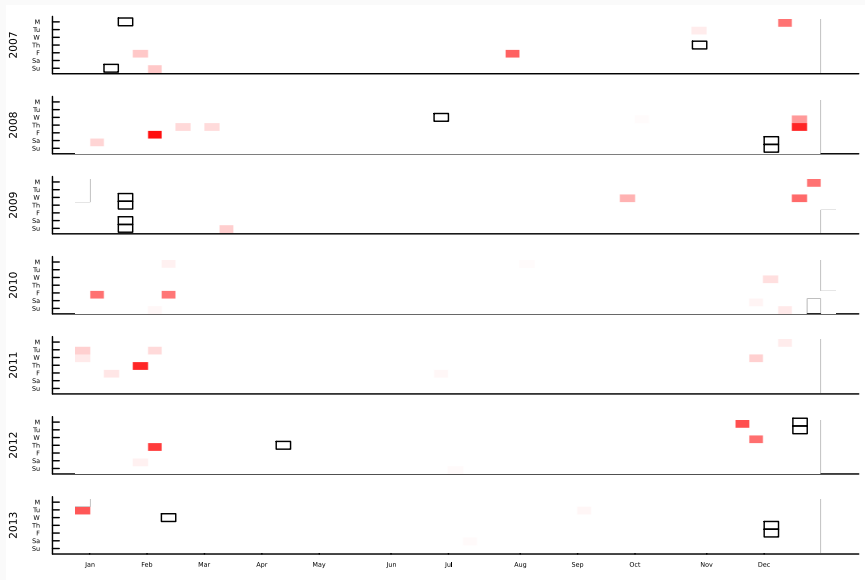
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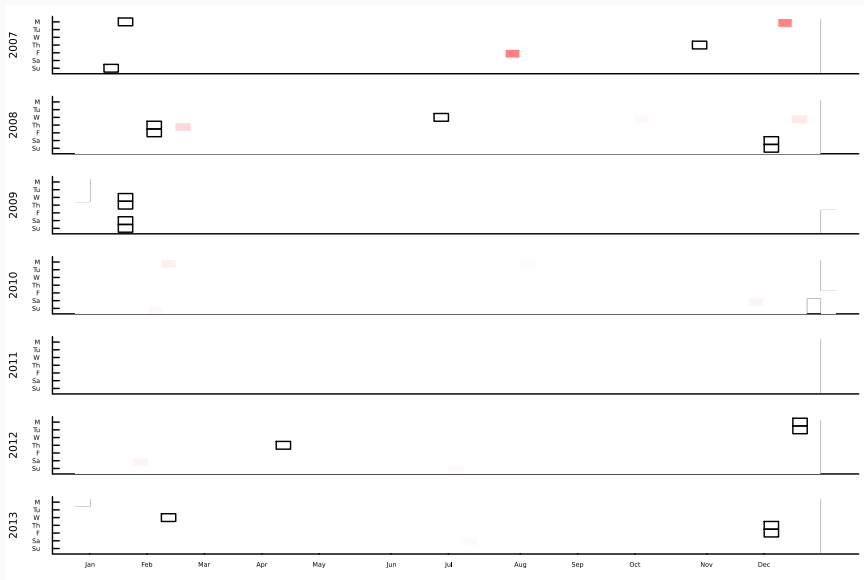
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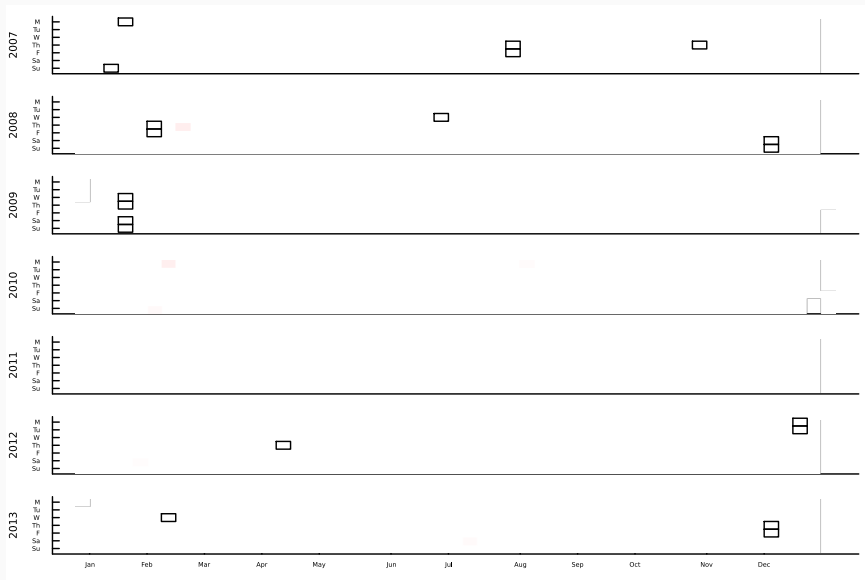


# Risk Period Iteration: discovering days to drive adequacy investment

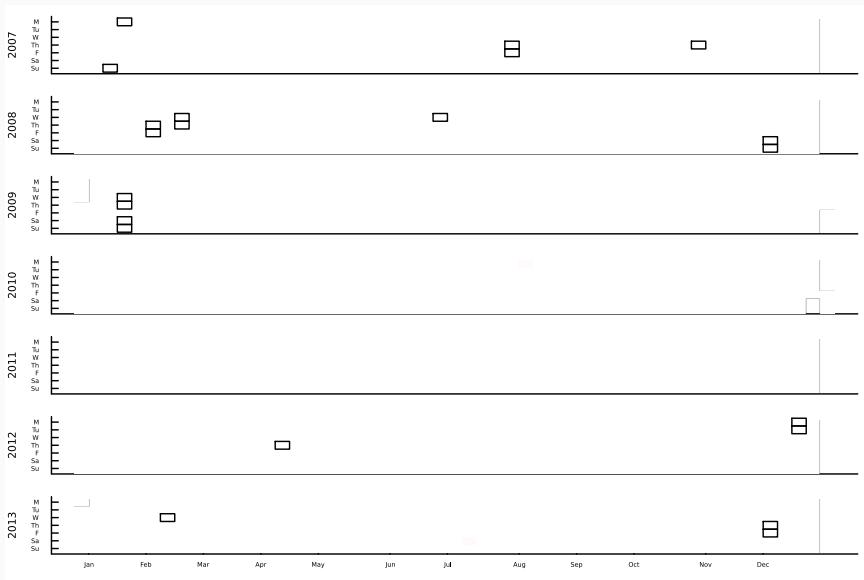




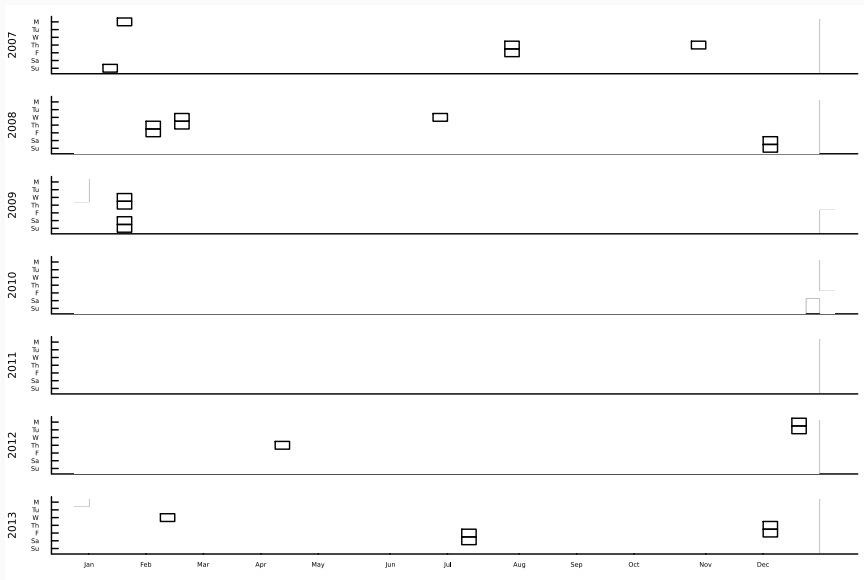
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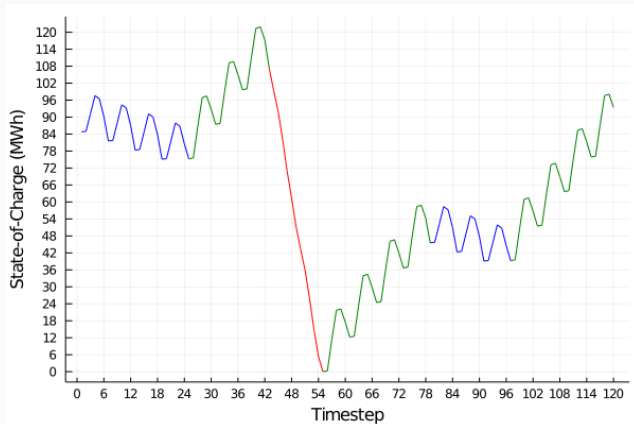


# Risk Period Iteration: discovering days to drive adequacy investment



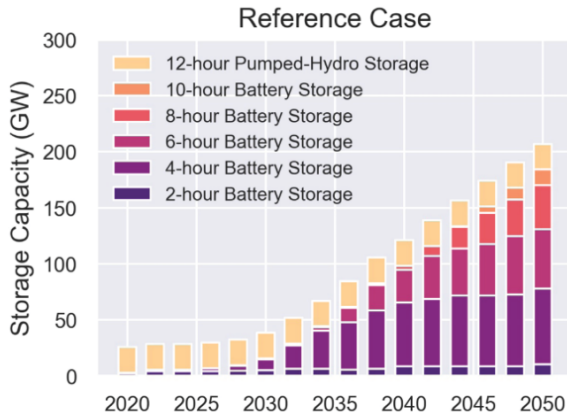
# Considering Energy Limited Resources

- Full “sparse” chronology repeats representative days in temporal sequences with minimal impact on problem size
- Risk periods can be explicitly represented within overall chronology



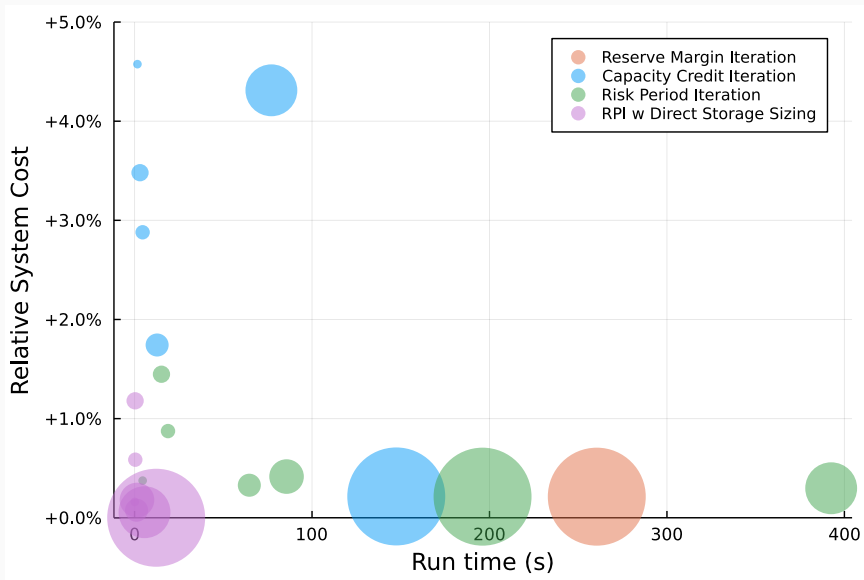
## Considering Energy Limited Resources

- Capacity credit-based frameworks require arbitrarily partitioning energy-limited technologies into subclasses
- With a direct risk period representation, resources can be sized flexibly to system needs



Blair, Nate, Chad Augustine, Wesley Cole, et al. 2022. *Storage Futures Study: Key Learnings for the Coming Decades*. Golden, CO: National Renewable Energy Laboratory. NREL/TP-7A40-81779. <https://www.nrel.gov/docs/fy22osti/81779.pdf>

# Adequacy Framework Performance Comparison (RTS-GMLC)



# Stay in touch!

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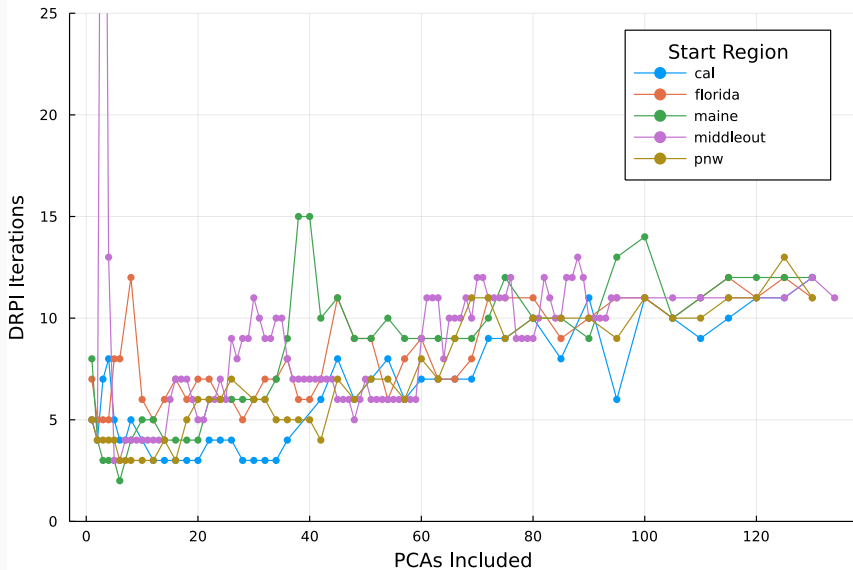
Gord Stephen  
gords@uw.edu

# Appendix

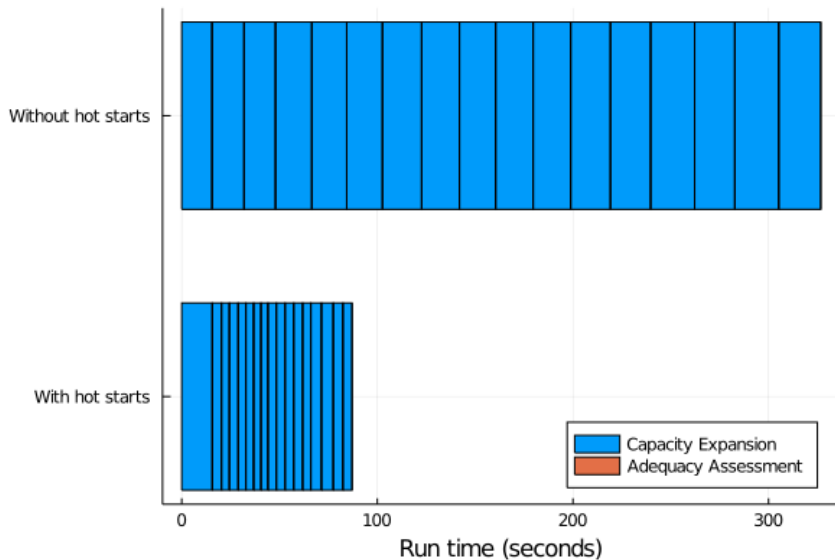
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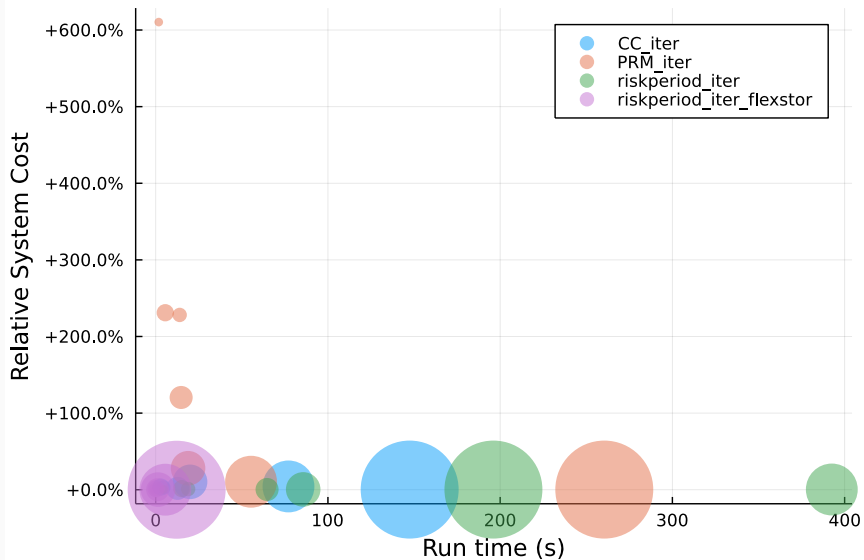
## Does Risk Period Iteration scale?



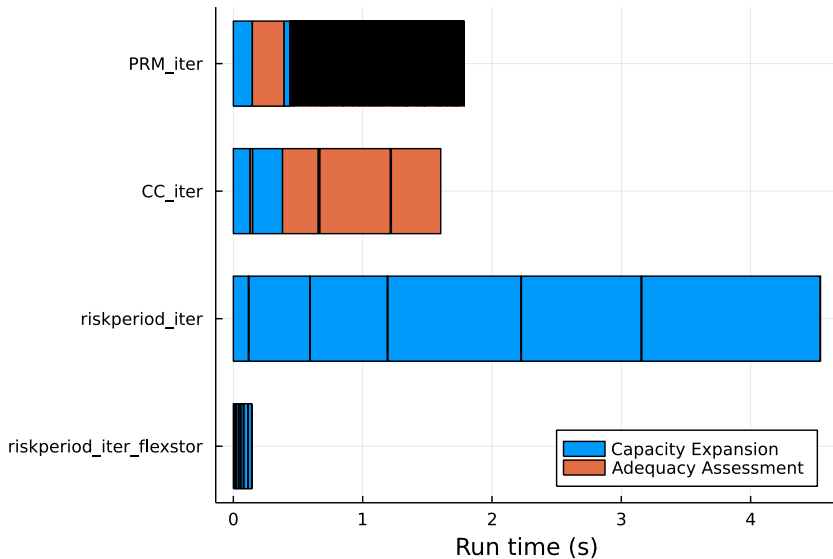
# Hot-starting Iterative Optimization Solves



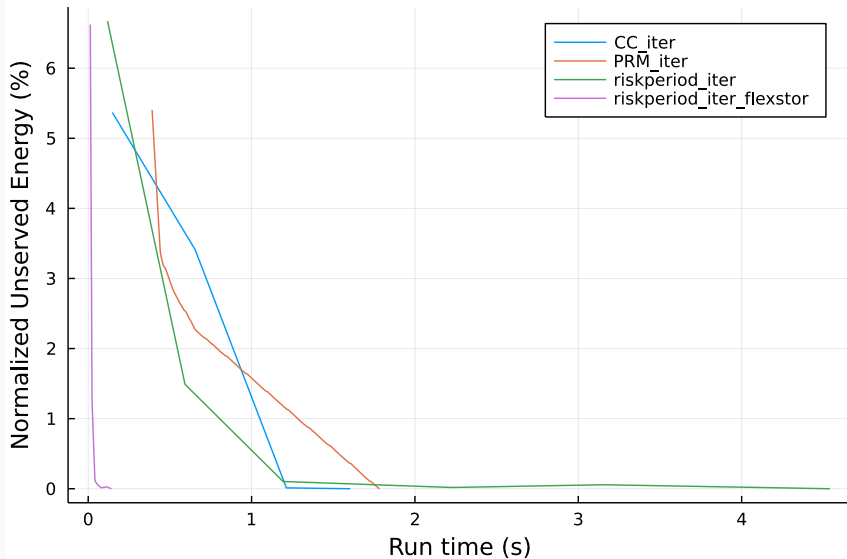
# Adequacy Framework Performance Comparison



# Adequacy Framework Performance Comparison



# Adequacy Framework Performance Comparison



## Sparse Chronological Energy Constraints (1/2)

$$\Delta e_d = \sum_{t \in T} p_{td} \quad \forall d \in D \quad (1)$$

$$\lfloor e \rfloor \leq \sum_{i=1..t} p_{id} \quad \forall t \in T, d \in D \quad (2)$$

$$\lceil e \rceil \geq \sum_{i=1..t} p_{id} \quad \forall t \in T, d \in D \quad (3)$$

## Sparse Chronological Energy Constraints (2/2)

$$e_{0,p+1} = e_{0,p} + N_p \Delta e_{d_p} \quad \forall p \in P \quad (4)$$

$$0 \leq e_{0p} + \lfloor e \rfloor_{d_p} \quad \forall p \in P \quad (5)$$

$$0 \leq e_{0p} + (N_p - 1) \Delta e_{d_p} + \lfloor e \rfloor_{d_p} \quad \forall p \in P \quad (6)$$

$$e_{0p} + \lceil e \rceil_{d_p} \leq E \quad \forall p \in P \quad (7)$$

$$e_{0p} + (N - 1) \Delta e_{d_p} + \lceil e \rceil_{d_p} \leq E \quad \forall p \in P \quad (8)$$

## Sparse Chronological Energy Constraints - Problem Size

- TD dispatch variables ( $p_{td}$ )
- D state-of-charge evolution variables ( $\Delta e_d$ )
- 2D state-of-charge bounding variables ( $\lceil e \rceil_d, \lfloor e \rfloor_d$ )
- D state of charge evolution definitional constraints (enforcing the definition of  $\Delta e_d$ )
- 2TD relative minima and maxima inequality constraints (enforcing the definitions of  $\lceil e \rceil_d$  and  $\lfloor e \rfloor_d$ )
- P boundary condition energy variables ( $e_{0p}$ )
- P boundary condition equality constraints (enforcing the definition of  $e_{0p}$ )
- 4P state of charge constraints