

Optimal dispatch and pricing for battery energy storage systems

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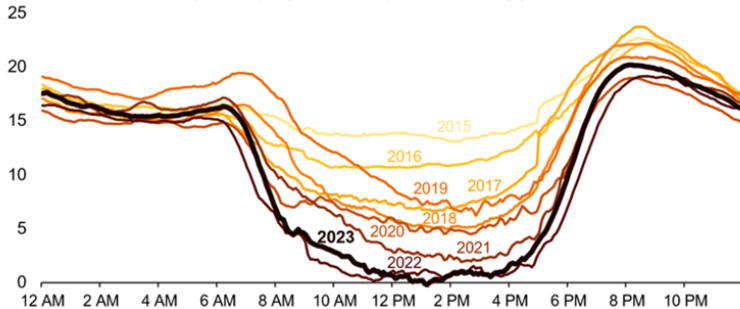
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As solar capacity grows, duck curves are getting deeper in California

California's duck curve is getting deeper

CAISO lowest net load day each spring (March–May, 2015–2023), gigawatts



Data source: [California Independent System Operator](#) (CAISO)

CAISO Duck curves [[California Independent System Operator](#)]



Rotohiko BESS (35 MWh, December, 2023) [<https://infratec.co.nz>]



Meridian's Ruakākā Battery Energy Storage System (BESS) is being officially opened in a ceremony later today.

Ruakaka BESS (100 MW, May, 2025) [<https://www.meridianenergy.co.nz>]



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Contact Energy Glenbrook BESS (100 MW, March 2026) [<https://contact.co.nz>]

Self dispatch versus central dispatch

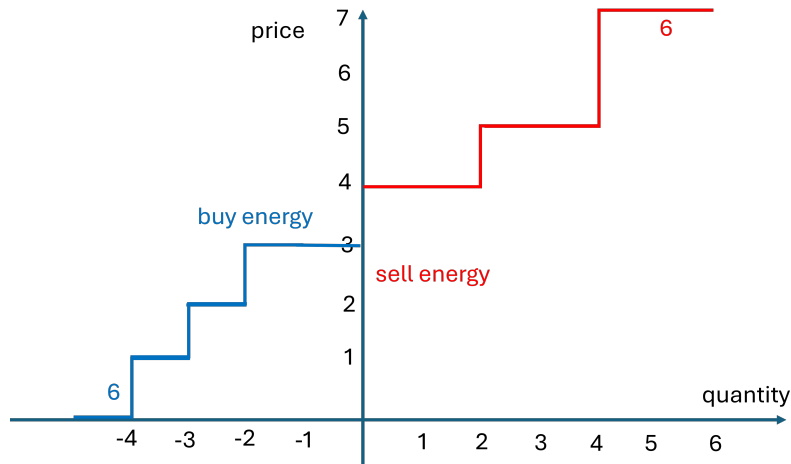
Self dispatch

- ▶ Battery forecasts/models prices and solves an optimization problem to maximize revenue from storage.
- ▶ System operator forecasts **exogenous** battery operation as part of net demand.

Central dispatch

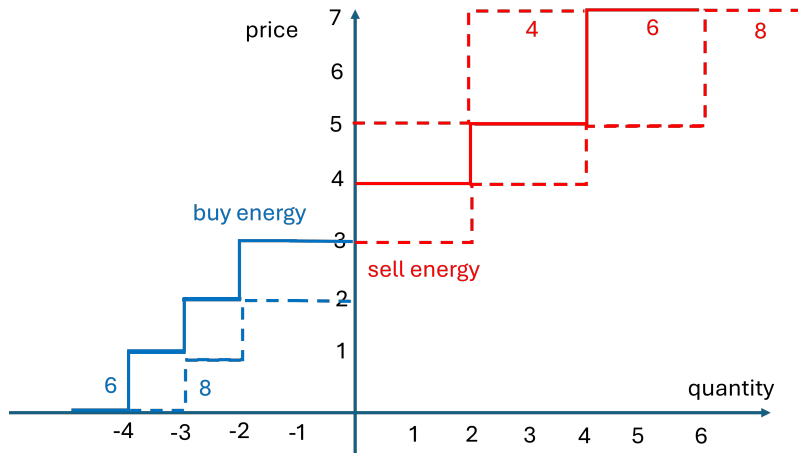
- ▶ Battery provides supply/demand curve defining what battery will sell/buy as price increases.
- ▶ System operator optimizes dispatch problem including **endogenous** battery operation.

Supply/demand offer curve for battery



Supply-demand curve for battery with capacity 10 and charge 6.

Supply/demand offer curves for battery



Curves depend on battery charge.

BESS central dispatch in New Zealand

- ▶ In NZ the system operator solves a dispatch problem every five minutes.
- ▶ Gate closure: Generators and BESS operators have offer curves for each five minute period, but cannot change offer curves within one hour of real time.
- ▶ Locational marginal energy prices are published in real time and affect the dispatch of plant and BESS.
- ▶ However, the settlement price of energy is the average of the six five-minute prices in each half hour.
- ▶ BESS operators are not paid the true value of arbitraging five minute prices.

Five minute gate closure?

- ▶ BESS operators currently lobbying for **five minute gate closure and settlement**.
- ▶ System operator needs offers locked in to stress test security of grid between each (security constrained) dispatch.
- ▶ Proposed compromise: maintain gate closure but enable **state-dependent offers** also known as **agent decision rules**.

Agent decision rules (ADRs)

- ▶ An **agent decision rule (ADR)** is a mapping from any known **parameter** of the stage t problem, and a 's state (storage) at end of t , to an energy offer in period t .
- ▶ An **agent Bellman function (ABF)** for agent a in period t is a **function** $W_a^t(y)$ that expresses the **expected future benefit** to a of being in **state** y at the end of period t .
- ▶ We can define an ADR for battery a using observed price $\pi(t)$ and its initial storage y_a and ABF W_a^t . Choose discharge u and charge v so as to:

$$\begin{aligned} \max_{u,v} \quad & \pi(t)(u - v) + W_a^t(y_a - u + \eta v) \\ \text{s.t.} \quad & 0 \leq y_a - u + \eta v \leq E_a \end{aligned}$$

Single-node electricity dispatch with ABFs

Given ABFs for battery agents $\mathcal{B}$, and state of charge $y(t-1)$:

$$\text{ADR}(t): \min \sum_{a \in \mathcal{G}} c_a x_a(t) + Lz(t) - \sum_{a \in \mathcal{B}} W_a^t(y_a)$$

$$\text{s.t.} \quad \sum_{a \in \mathcal{G}} x_a(t) + \sum_{a \in \mathcal{B}} u_a(t) - \sum_{a \in \mathcal{B}} v_a(t) + z(t) = d(t), \quad [\pi(t)]$$

$$x_a(t) \in \mathcal{X}_a(x_a(t-1)), \quad a \in \mathcal{G},$$

$$(y_a(t), u_a(t), v_a(t)) \in \mathcal{Y}_a(y_a(t-1)), \quad a \in \mathcal{B},$$

$$z(t) \in [0, d(t)].$$

New dispatch process with ADRs

- ▶ Generator agents provide system operator with marginal costs.
- ▶ Battery agent a provides system operator with ADR defined by increasing concave ABF W_a^t .
- ▶ System operator solves single-stage problem $\text{ADR}(t)$ and computes dispatch and system marginal price $\pi(t)/\text{MWh}$.
- ▶ Generator is paid $\pi(t)$ per MWh
- ▶ Battery is paid $\pi(t)(\text{charge} - \text{discharge})$.

Remarks

- ▶ $\text{ADR}(t)$ is a deterministic convex optimization problem (assuming no unit commitment).
- ▶ This means price $\pi(t)$ gives **budget balance** for system operator (i.e. revenue adequacy).
- ▶ Price $\pi(t)$ defines a perfectly competitive equilibrium for stage t , so **agents recover costs**.
- ▶ Does dispatch problem $\text{ADR}(t)$ yield **social optimum**?

What can go wrong?

- ▶ In $\text{ADR}(t)$, dispatch $x_a^*(m)$ will vary with $\pi^*(m)$.
- ▶ But BESS **offer in market** is an increasing function of $\pi^*(m)$.
- ▶ BESS offers use an ABF $W_a^t(y)$ that depends on the state of charge y at the end of period.
- ▶ $W_a(y)$ depends on agent's beliefs of future prices.
- ▶ If agent believes future prices are conditioned on current price π^* then **$W_a^t(y)$ should be $W_a^t(y, \pi^*)$** .
- ▶ We want $x_a^*(n)$ to solve

$$\max_{x_a \in \mathcal{X}_a(x^*(n-))} \{ \pi^*(n)^\top x_a - c_a^n(x_a) + W_a^t(x_a, \pi^*(n)) \};$$

- ▶ $x_a^*(n)$ can be **decreasing in $\pi^*(n)$** .

Social optimality

- ▶ Can ADR offers give a socially optimal solution?
- ▶ Assume:
 - ▶ All agents and SO have the same probability distribution of future events;
 - ▶ Each agent knows all system parameters including marginal costs of other agents;
 - ▶ Each agent can compute the social future cost function C^t .
 - ▶ $-\partial C^t = \Pi_{a \in \mathcal{G} \cup \mathcal{B}} \partial W_a^t$, i.e., $-C^t$ is **gradient separable** by agent.
- ▶ If agent a uses ABF W_a with $\frac{\partial W_a}{\partial y_a} = -\frac{\partial C^t}{\partial y_a}$ then this gives the same solution as social optimum.
- ▶ The optimal agent **generation might not be increasing in price.**

Restrictions to ensure monotonicity for dispatch problem

- ▶ Can we **iron** optimal offers to make them nondecreasing?
- ▶ Yes, but ...
- ▶ ... a dispatch problem with

$$C^t(x, \pi) = - \sum_{a \in \mathcal{G} \cup \mathcal{B}} W_a(x_a, \pi)$$

where π is the system marginal price is **not convex**, and can yield multiple dispatch solutions.

- ▶ So instead we restrict the form of ADRs to only allow agent Bellman functions that preclude decreasing offers.
- ▶ Example: ABFs that depend only on x_a .
- ▶ Restrictions sacrifice some social optimality.

Example: one battery, one ramping generator

$x(t)$ = dispatch of generator in period t ;

$\bar{x}$ = dispatch of generator in period $t - 1$;

$y(t)$ = storage in battery at end of period t ;

$\bar{y}$ = storage in battery at end of period $t - 1$;

u = discharge from battery in period t ;

v = charge input to battery in period t ;

$$\mathcal{X}(\bar{x}) = \{x \mid 0 \leq x \leq q, x - \bar{x} \leq \rho, \bar{x} - x \leq \sigma\},$$

$$\mathcal{Y}(\bar{y}) = \{(y, u, v) \mid 0 \leq y \leq E, 0 \leq u \leq r, 0 \leq v \leq s, \\ y = \bar{y} - u + \eta v\}.$$

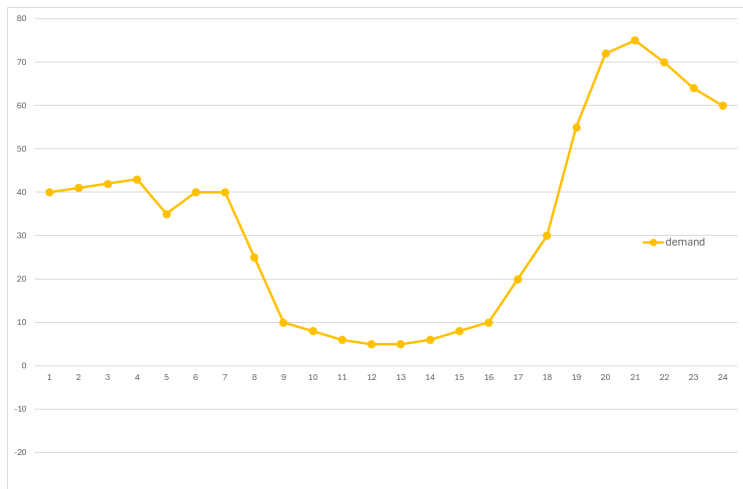
An example: one battery, one ramping generator

Assume $T = 24$, $c(x) = 70.0x$, $\sigma = \infty$. Other parameters are as follows.

$q = 70.0$	$E = 30.0$	$\eta = 0.8$
$r = 10.0$	$s = 10.0$	$\rho = 10.0$
$L = 500.0$	$x^0 = 35.0$	$y^0 = 4.0$

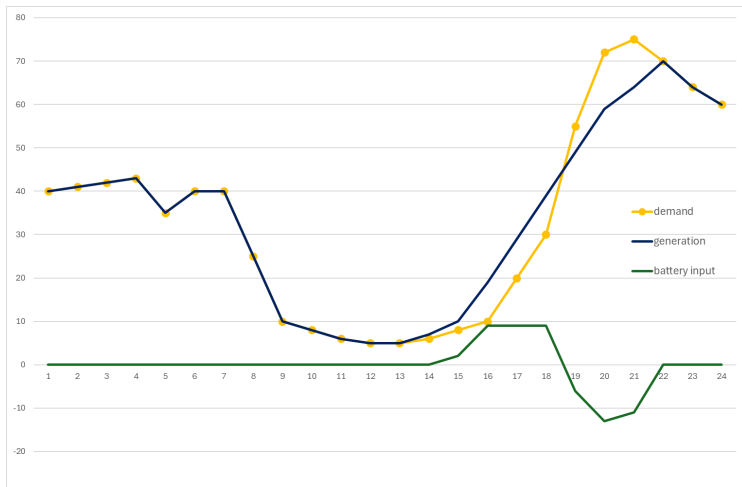
Parameter values for example

Example demand



Example values of $d(t)$ for $t = 1, 2, \dots, 24$. We add stagewise independent random noise chosen from $-4.0, -2.0, 0.0, 2.0, 4.0$ with equal probability

Deterministic solution



Zero-noise solution showing generation x superimposed on demand, and battery net input $v - u$ for $t = 1, 2, \dots, 24$.

SDP versus ADR with noise

- ▶ Solve SDP with independent noise. Expected cost = \$52,377, and optimal future cost function $C^t(x, y)$, $t = 1, \dots, 24$.
- ▶ Let $(\tilde{x}^t, \tilde{y}^t)$ denote values of generation and values of battery storage at each stage of zero-noise solution.
- ▶ Suppose ABF is a separable approximation of C^t :

$$W_a^t(x) = -C^t(x, \tilde{y}^t), \quad a \in \mathcal{G};$$

$$W_a^t(y) = -C^t(\tilde{x}^t, y), \quad a \in \mathcal{B}.$$

- ▶ Simulated policy gives \$52,688 $\pm$ 55. Some social optimality is lost since $(\tilde{x}^t, \tilde{y}^t) \neq (x^*(t), y^*(t))$ (varying with each sample path).

Conclusions

- ▶ Under **restrictive assumptions** there is a set of ADRs for each agent that maximizes expected welfare.
- ▶ In practice, social optimality is compromised by **lack of gradient separability**, **lack of information**, or **decreasing generation with price**.
- ▶ ADRs can encompass **different beliefs**.
- ▶ System operator does not influence future prices through imperfect forecasts of demand - this falls on aggregating agents' beliefs through ADRs.
- ▶ Changes required to update dispatch software are relatively simple, but offer data must be constrained to ensure convexity.

The End

Any questions?

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For the paper go to `https://www.epoc.org.nz/papers/ADRPaperForEPOC.pdf`