

The background of the slide is a high-angle aerial photograph of a coastline. The water is a deep teal color, with white-capped waves crashing against a rocky shore. The rocks are light-colored and vary in size. A dark rectangular overlay is positioned in the lower-left quadrant of the image.

# Battery optimisation

MAST30034 Applied Data Science



# Agenda

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overview

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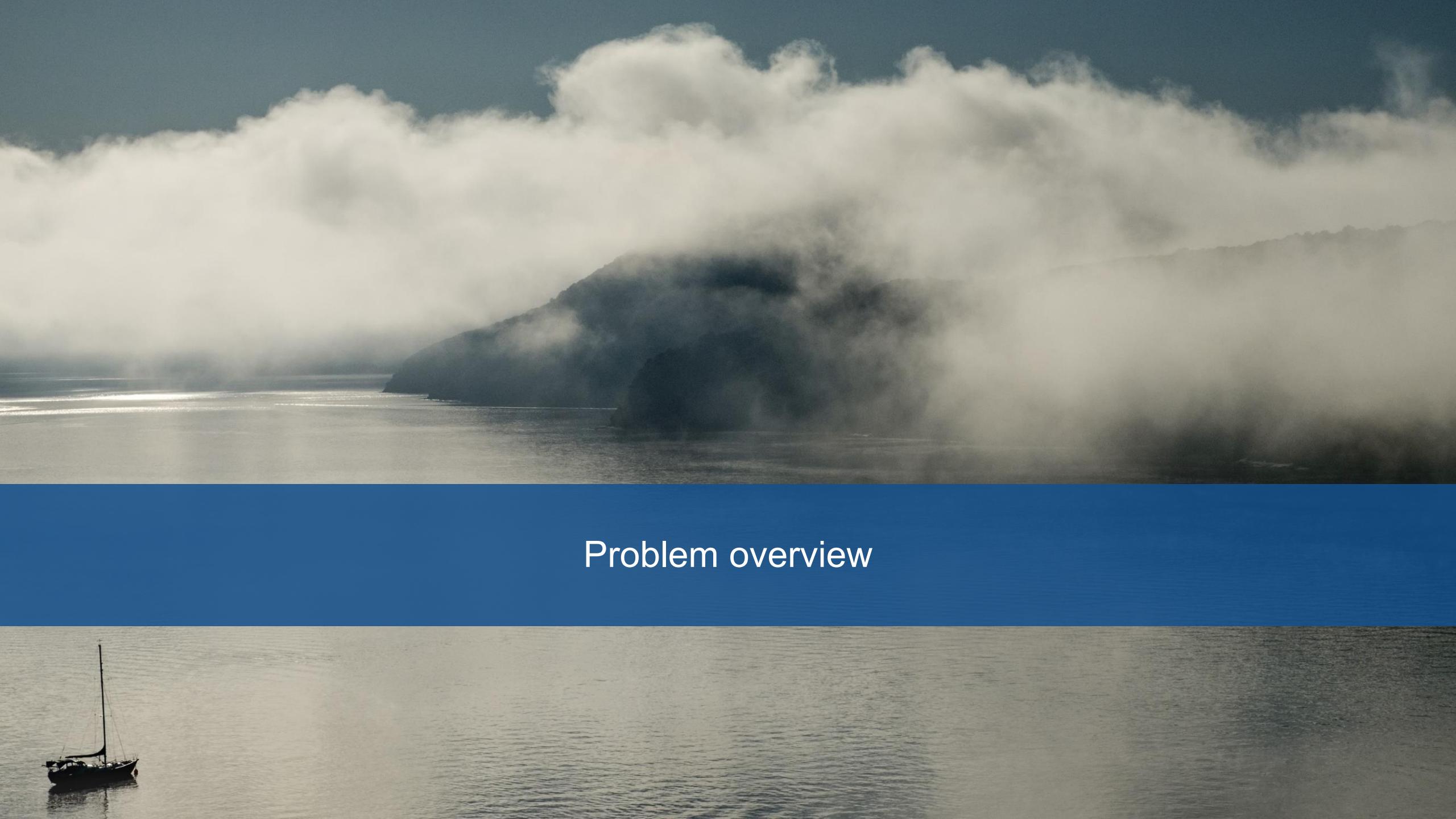
Energy  
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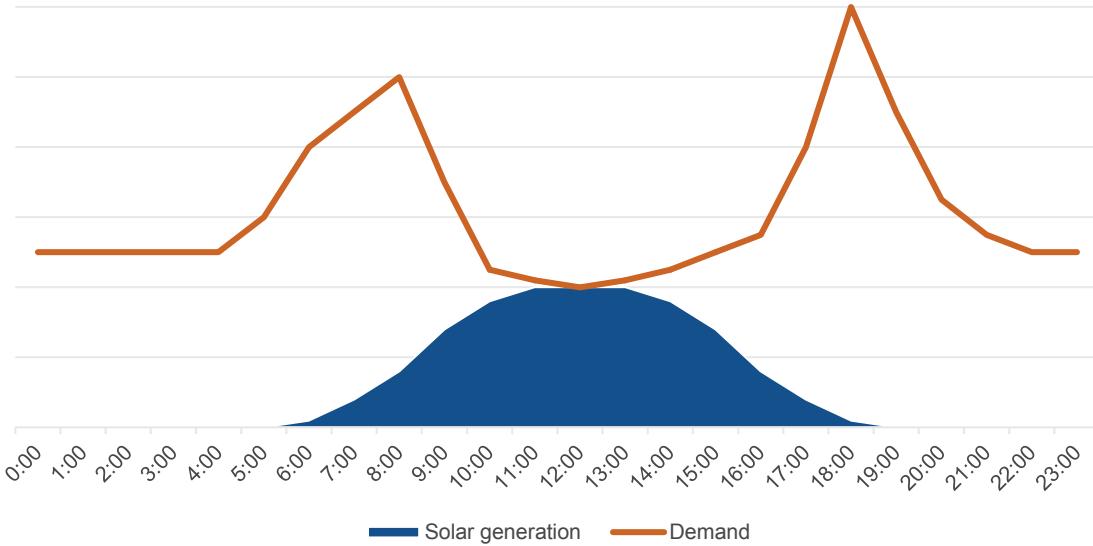
Evaluation  
process



## Problem overview

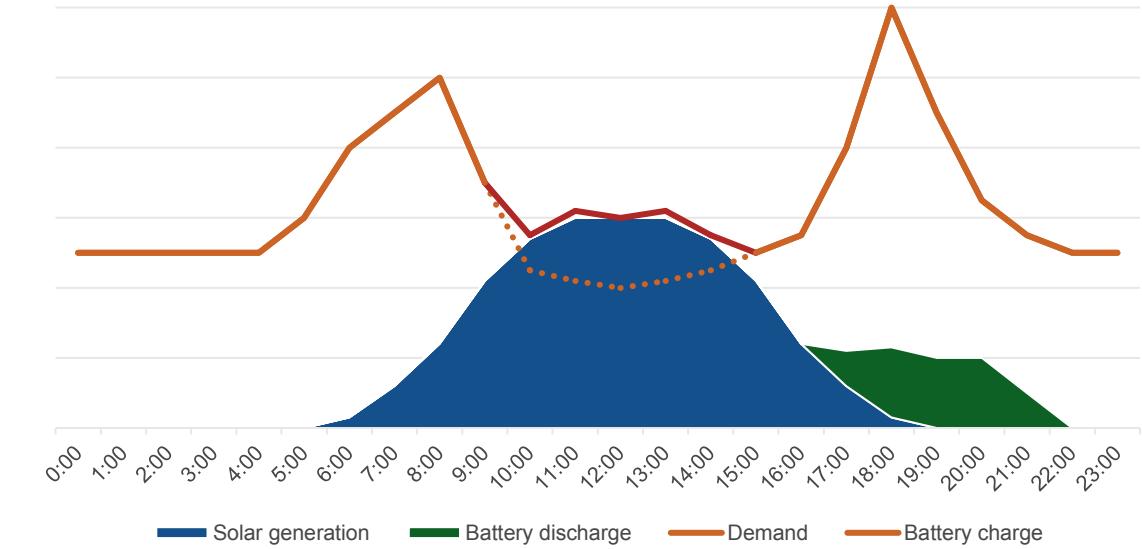
# Introducing batteries

## Without battery



- In Australia, grid electricity demand has a bath-tub shape (called the duck-curve) due to rooftop solar on homes and commercial properties reducing the amount of electricity needed during the middle of the day.
- Additional solar generation is therefore limited in its ability to meet demand, despite having low capital costs and no short-run marginal cost (no fuel – it is free to operate).

## With battery



- Batteries add energy demand when they charge, and then add to energy supply when they discharge.
- Energy storage technologies, such as batteries, are critical for the renewable energy transition due to their ability to shift energy to low sunlight/wind periods.
- Note the large solar capacity (compared to without battery) due to the battery 'unlocking more generation'. This displaces more expensive and high-emitting generation.

# The problem

## Objective: Mandatory task

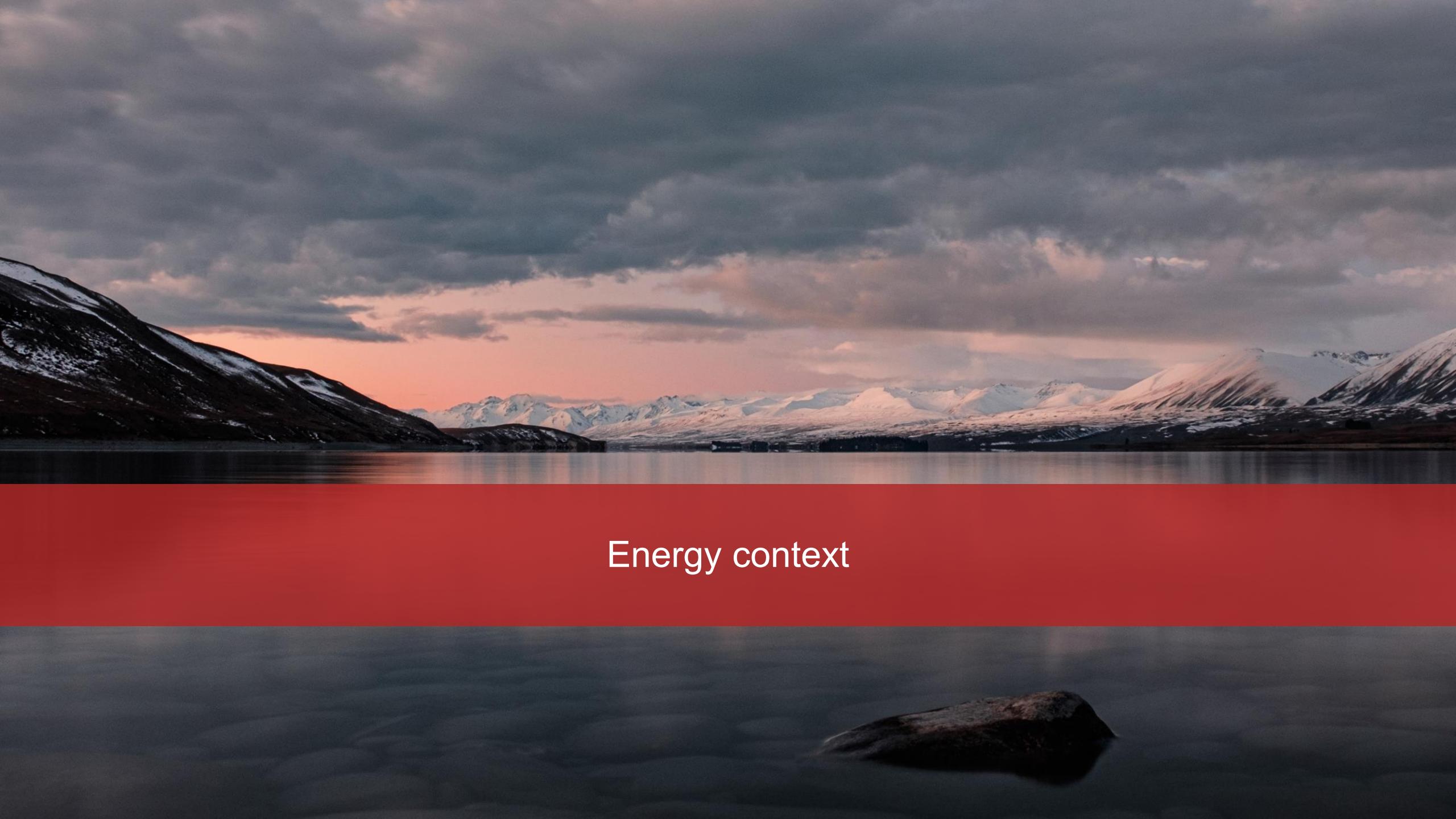
- Develop an algorithm that determines the optimal charge and discharge behavior of a battery based in Victoria.
- Maximise revenues by charging when electricity prices are low and discharging when prices are high.

## Objective: Bonus task

- Inform current charging behavior without using future prices.

## Industry purpose: Revenue maximisation for...

- Stand-alone grid batteries
- Residential battery aggregators (Virtual Power Plants)
- Renewable energy developers



Energy context

# National Electricity Market (NEM)

## Overview of NEM

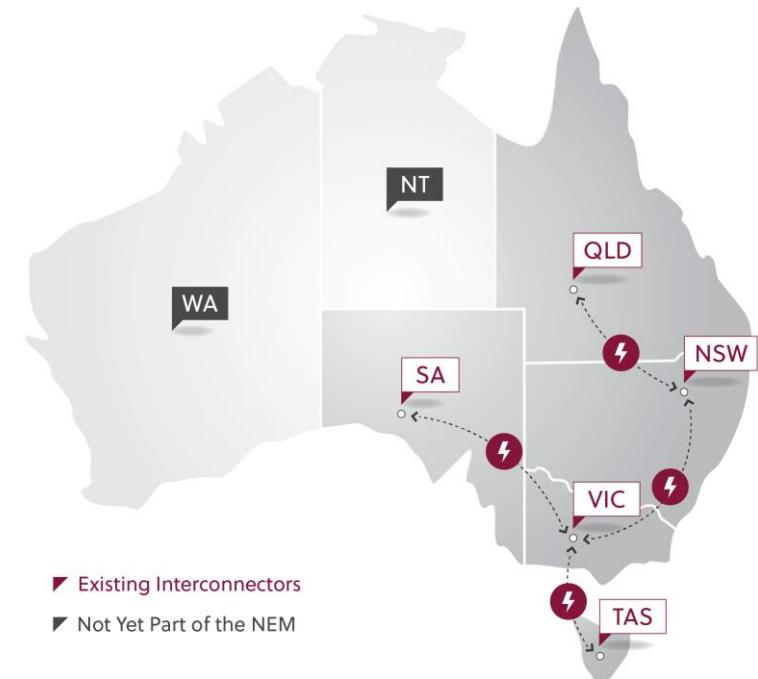
- Comprises 5 states (ACT is considered to be a part of NSW)
- Interconnectors between the states allow for the sharing and trading of electricity (up to certain specified limits)

## Price setting

- Prices are set every 30 minutes based on prevailing supply and demand (called the spot price)
- Each state has its own price

## Generation

- Every generator is registered to a single state
- Dispatchable generators include coal, gas, biomass and hydro
- Intermittent generators include wind and solar
- Storage technology includes pumped hydro and batteries



# Price setting in the NEM: 30-minute spot price

As an illustrative example, consider a closed system with a 10MW (pronounced megawatt) coal generator, a 10MW solar farm, and a 10MW gas generator.

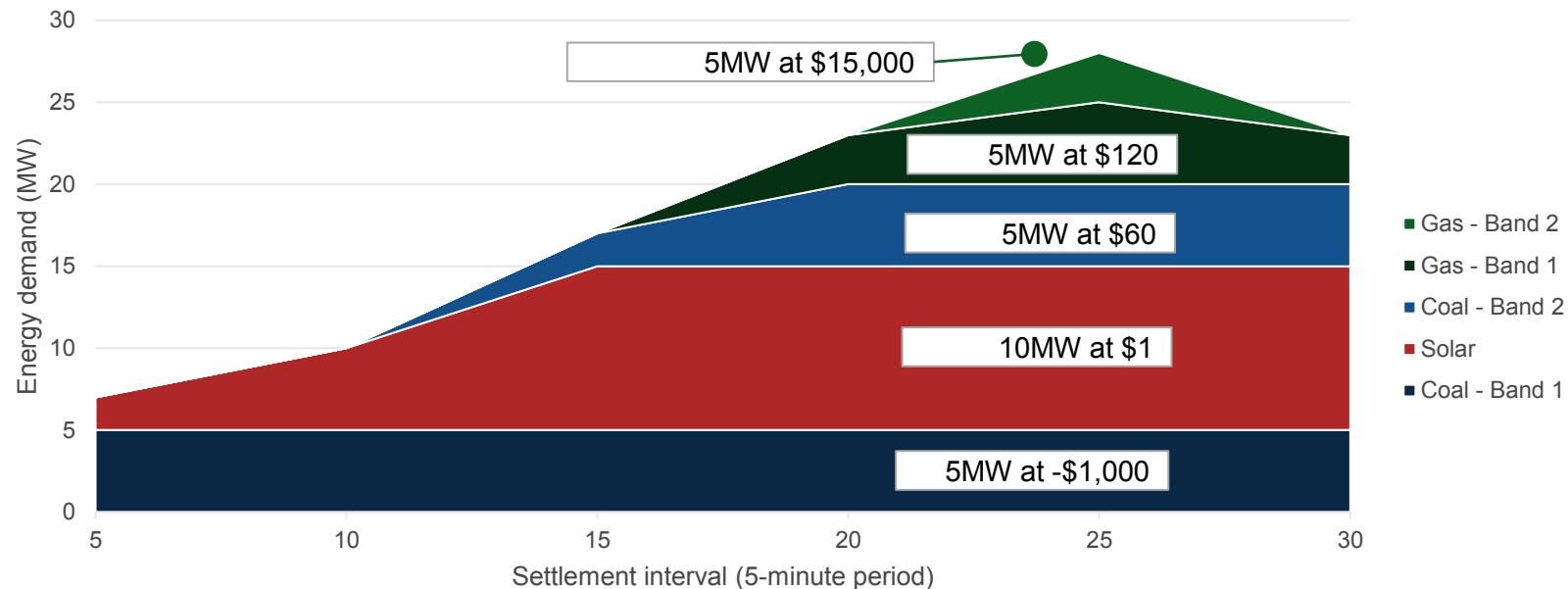
**The coal generator bids 5MW at -\$1,000:** It is prohibitively expensive to restart a coal turbine and causes significant wear-and-tear to run it discontinuously, so coal generators tend to put their minimum stable volume at the market floor price to guarantee dispatch.

**The coal generator bids its remaining 5MW at \$60:** The short-run marginal cost for coal is currently around \$30-\$50 per MW so this ensures profitability for ramping above minimum stable levels.

**The solar farm bids all its 10MW at \$1:** Solar is a price-taker and has no short-run marginal cost to operate (no fuel required).

**The gas generator bids 5MW at \$120:** The short-run marginal cost for gas is currently around \$80-\$120 per MW and so this price ensures it switches on when profitable. Unlike coal, some gas turbines can switch on and off.

**The gas generator bids its remaining 5MW at \$15,000:** When demand reaches extreme levels, there is no competition and so generators can bid at the market-cap.



In this simplified example, the following settlements occur:

- In the 5-minute and 10-minute periods, demand is low and is serviced by the lowest cost generation. The price is \$1 in both periods and this is received by both solar and coal for their respective volumes dispatched.
- In the 15-minute period, demand rises and coal ramps up to meet this. The price is set at \$60 which is received by both coal and solar for their respective volumes dispatched.
- In the 20-minute and 30-minute periods, demand is near its peak and gas switches on to ensure adequate supply. The price is set at \$120 and this is received by all generators for the respective volumes dispatched.
- In the 25-minute period, demand momentarily reaches its peak. The price is set at \$15,000 and all generators receive this price for their respective generation.

**The price for the 30-minute period is set at the average of the 6 periods = ~\$2,550/MWh. All generators receive this price for their generation output during the 30-minute period.**

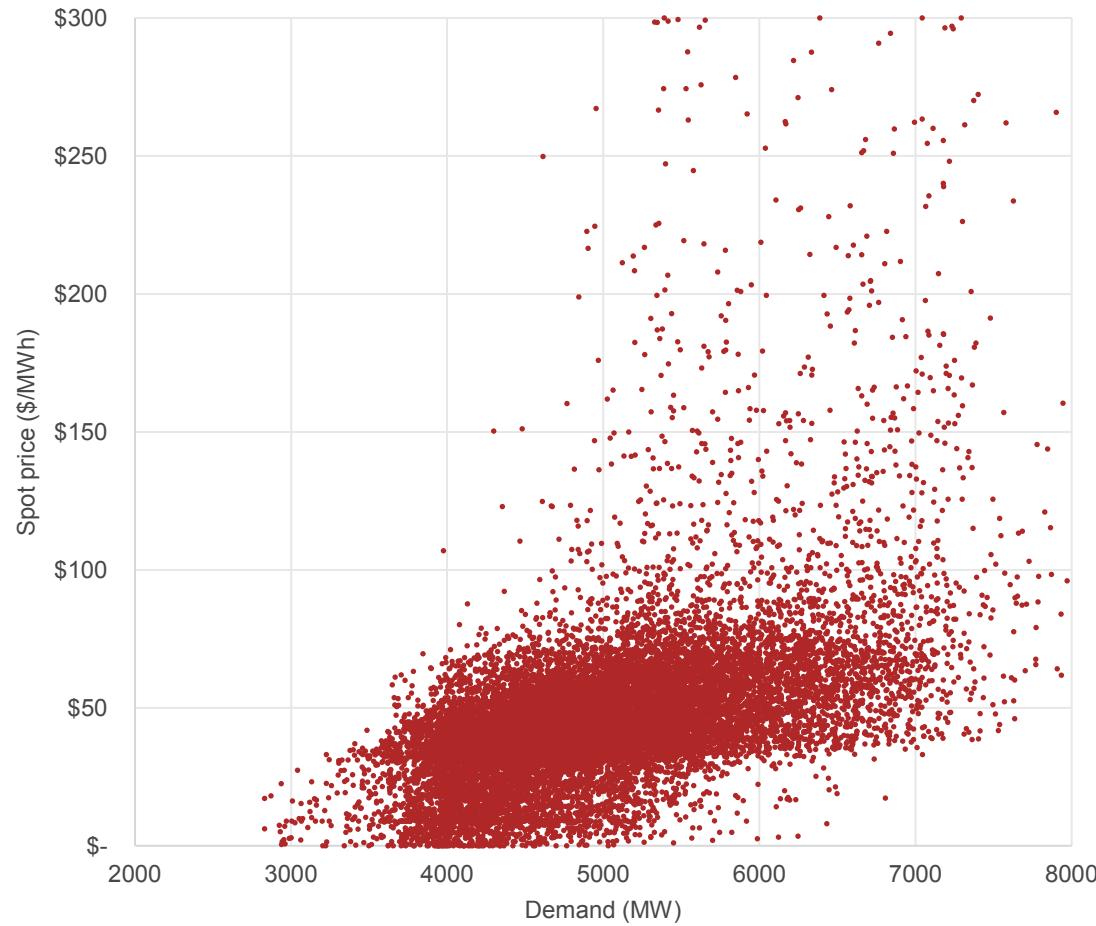
# A note on units



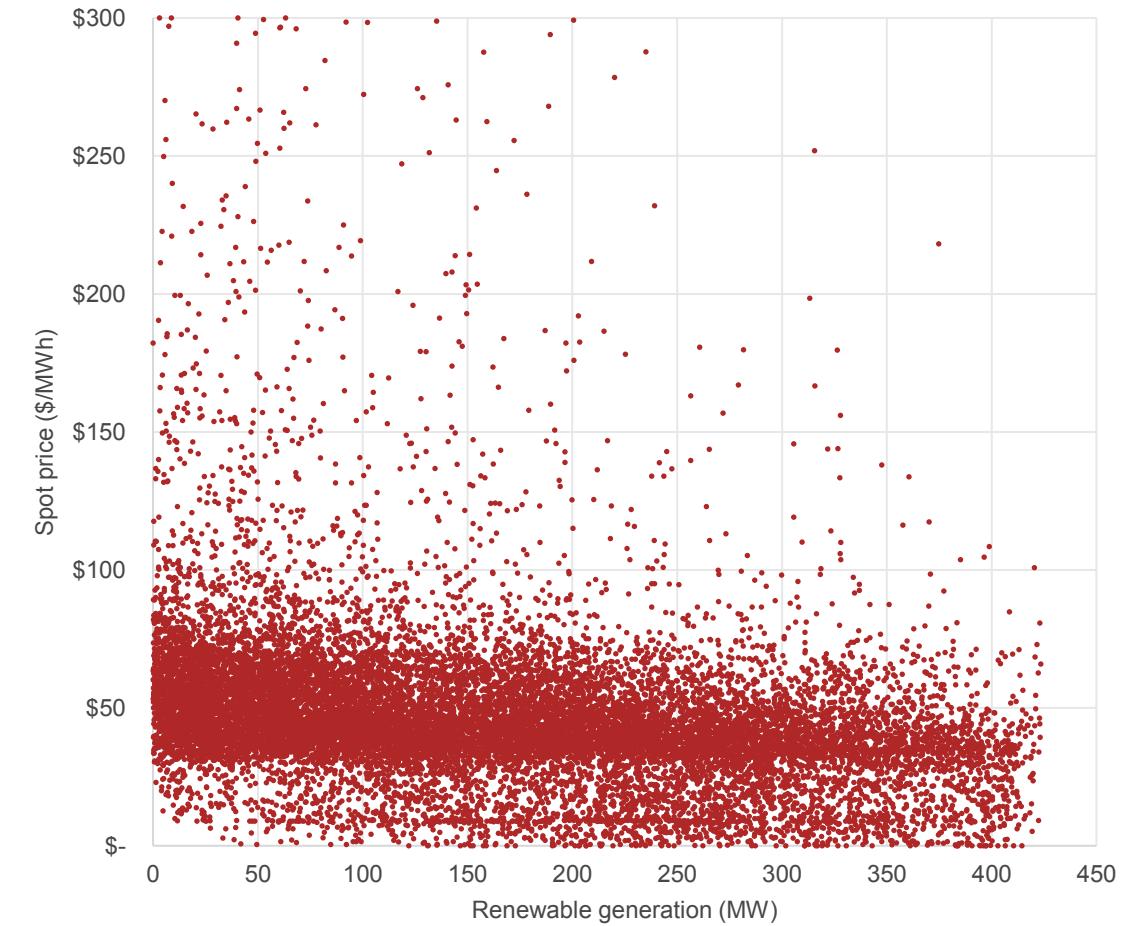
12 MW	1 hour	12 MWh
12 MW	30 minutes	6 MWh
12 MW	5 minutes	1 MWh
60 MW	1 hour	60 MWh
60 MW	30 minutes	30 MWh
60 MW	5 minutes	5 MWh

# Electricity price trends: Victorian spot price (2020)

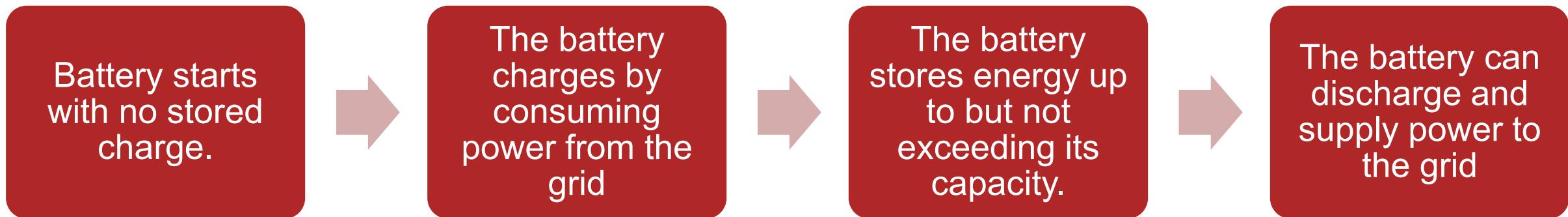
**Spot price vs. demand**



**Spot price vs. renewable generation**



# How batteries function





## Problem description



# Problem objective

## Mandatory task: The perfect but impractical battery

- Develop an algorithm to maximise the revenues of a grid connected battery assuming perfect future price visibility. VIC price is the only variable required for this optimisation.

## Bonus task: The practical but imperfect battery

- Develop another algorithm to maximise the revenues of a grid connected battery assuming no future visibility of price. The usage of renewables and demand data available at each period is allowed given the ability to forecast these.

## Requirements for both tasks

- Ensure all technical and operational criteria are met.
- Use the assigned Training and Cross-Validation periods to develop the algorithm(s).
- Output revenues and battery dispatch from the Test period for comparison with other groups.

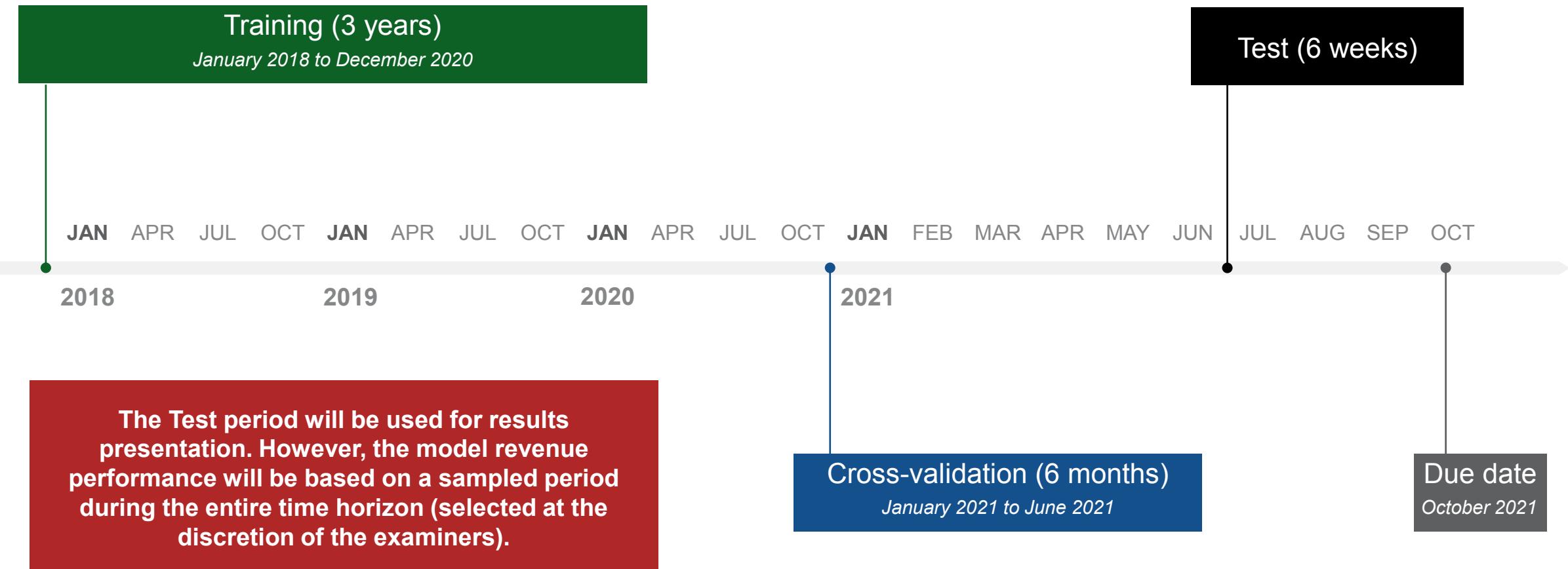
# Data points enabled for use at time t in the models

Data point	Examples	Mandatory task	Bonus task
Prices before current period	$Price_{t-n}$	Allowed	Allowed
Price during current period	$Price_t$	Allowed	Allowed
Price after current period	$Price_{t+n}$	Allowed	Not allowed
Demand/supply before current period	$Variable_{t-n}$	Allowed but not required	Allowed
Demand/supply during current period	$Variable_t$	Allowed but not required	Allowed
Demand/supply after current period	$Variable_{t+n}$	Allowed but not required	Allowed

n is an element of [1, 48] half-hour periods

# Data timeline for battery calibration

To ensure comparable outcomes across the groups, identical time periods should be used for Training, CV and Testing.



# Interim deliverables (check-points)

Week	Week starting	Deliverable	Objective
1	30 August	How much energy is consumed in 10 minutes at 90MW?	Demonstrate an understanding of electricity fundamentals.
2	6 September	What is the VIC spot price on 01/07/2020 15:00?	Demonstrate an understanding of the electricity data provided.
3	13 September	What is the maximum revenue a battery can make on 17/07/2020 assuming it starts the day discharged and can only discharge 580MWh (the battery has a single charge cycle for this test)?	Demonstrate an understanding of battery revenue maximisation.
4	20 September	Midsemester break	
5	27 September	Provide a proposed approach on how you will optimise revenues with your final algorithm (200 words)	Demonstrate that you are able to meet the final deadline. Students may elect to use a different approach for final solution.
6	04 October	None	

# Battery technical and commercial properties

Variable	Unit	Value	Description	Data source
State	Text	VIC	The regional node determines the spot market pricing.	-
Battery power	MW	300	Maximum instantaneous rate of energy release or energy charging.	Proposed battery for Deer Park, Melbourne <sup>1</sup> .
Battery capacity	MWh	580	Maximum energy that can be stored in the battery at full charge.	Proposed battery for Deer Park, Melbourne <sup>1</sup> .
Charge efficiency	%	90	Electricity to chemical conversion rate into stored battery capacity.	Australian Energy Market Operator <sup>2</sup>
Discharge efficiency	%	90	Chemical to electrical conversion rate into grid dispatched energy.	Australian Energy Market Operator <sup>2</sup>
Marginal Loss Factor	#	0.991	Losses associated with energy transmission.	Laverton North as shadow connection.
Capital cost	\$	0	Assume all development and procurement capital is sunk.	-
Fixed O&M	\$/kW/year	8.10	Costs associated with fixed operations and maintenance.	Australian Energy Market Operator <sup>3</sup>
Variable O&M	\$/MWh	0.00	Costs associated with variable operations and maintenance.	Australian Energy Market Operator <sup>3</sup>

1. <https://www.afr.com/companies/energy/transgrid-to-install-big-battery-for-melbourne-20210705-p586wv>

2. [https://www.aemo.com.au/-/media/Electricity/NEM/Planning\\_and\\_Forecasting/Inputs-Assumptions-Methodologies/2019/2019-Projections-for-Small-Scale-Embedded-Technologies-Report-by-CSIRO.pdf](https://www.aemo.com.au/-/media/Electricity/NEM/Planning_and_Forecasting/Inputs-Assumptions-Methodologies/2019/2019-Projections-for-Small-Scale-Embedded-Technologies-Report-by-CSIRO.pdf)

3. <https://aemo.com.au/en/energy-systems/major-publications/integrated-system-plan-ispl/2022-integrated-system-plan-ispl/current-inputs-assumptions-and-scenarios>

# Illustrative algorithms for consideration

Algorithm	Description	Revenue maximisation	Practicality (for Bonus task)
Fixed charge/discharge levels <i>(Provided as illustrative example 1)</i>	A battery that charges below a set price level, and discharges above a set price level.	Poor	High
Time of day charge/discharge <i>(Provided as illustrative example 2)</i>	A battery that charges and discharges at fixed periods in the day. Can be adjusted to have different periods based on season/month/day-of week etc.	Poor	High
Look ahead charge/discharge <i>(Provided as illustrative example 3)</i>	A battery that compares current prices to future observed prices to determine whether to charge or discharge.	Moderate	Poor

# Data sources for problem (all sourced from AEMO)

## Mandatory task

- Spot price data for Victoria  
Training and CV period (01/01/2018-30/06/2021)  
Illustrative example provided with problem
- Spot price data for Victoria  
Test period (01/07/2021-11/08/2021) – More data provided for t+n functions  
<https://aemo.com.au/energy-systems/electricity/national-electricity-market-nem/data-nem/aggregated-data>

## Bonus task

- Operational demand data for all 5 states
- Available dispatchable generation for all 5 states
- Available intermittent (renewable) generation for all 5 states
- Interconnector limits between all 5 states



# Evaluation process

# Mandatory task (7%)

Grade	Technical performance Battery operates within technical parameters (slide 17)	Revenue performance With comparison to the best performing group	Code quality Approach, efficiency and readability
7%	Correct implementation <small>(Trivial errors that don't impact revenues are allowed)</small>	Within X% of best performing group's revenue	Fair to Good
6%		Within Y% of best performing group's revenue	Poor
5%		Within Y% of best performing group's revenue	Fair to Good
4%		Within Y% of best performing group's revenue	Poor
3%		Better than baseline models provided	Fair to Good
2%		Better than baseline models provided	Poor
1%		Less or equal to than baseline models provided	Fair to Good
0%	Incorrect implementation	Less or equal to than baseline models provided	Not assessed

X and Y will be based on the number of groups and the distribution of results

# Bonus task (3%)

Grade	Proposed approach Description of adjustments to switch to model with no price foresight	Commentary on challenges Explanation on the functional and practical implications of no price foresight
3%	Credible approach	Excellent commentary
2%	Poor approach	Fair or poor commentary
1%	Not attempted	Fair or poor commentary
0%	Not attempted	Not attempted