



THE UNIVERSITY
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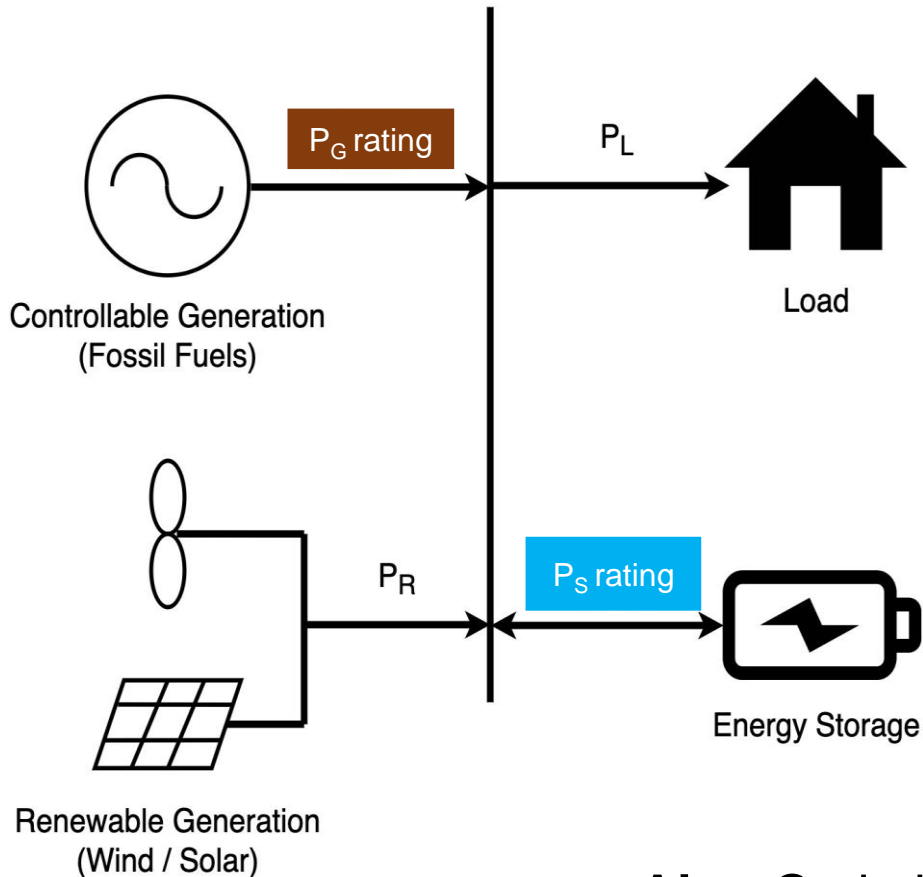
ENERGY STORAGE REQUIREMENTS FOR THE SA GRID

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Outline

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 - **Why Genetic Algorithms?**
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Project Aim



- P_G : Controllable Generation
- P_R : Renewable Generation
- P_S : Energy Storage
- P_L : Total Load

$$P_G + P_R + P_S = P_L$$

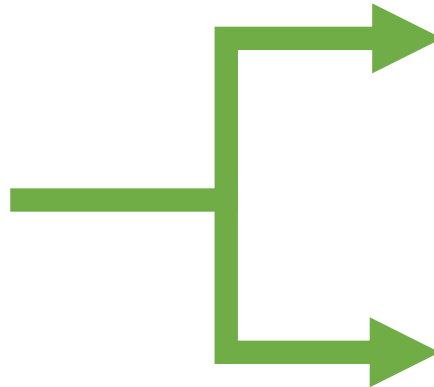
Aim: Optimise P_S rating to minimise P_G rating

Background

- **Energy Transition:** Fossil → Renewable
- Meet 'Renewable Energy Target' scheme
 - Reduce GHG emissions
 - Generate more electricity from renewables



Non - renewables



Renewables
(Wind / Solar)



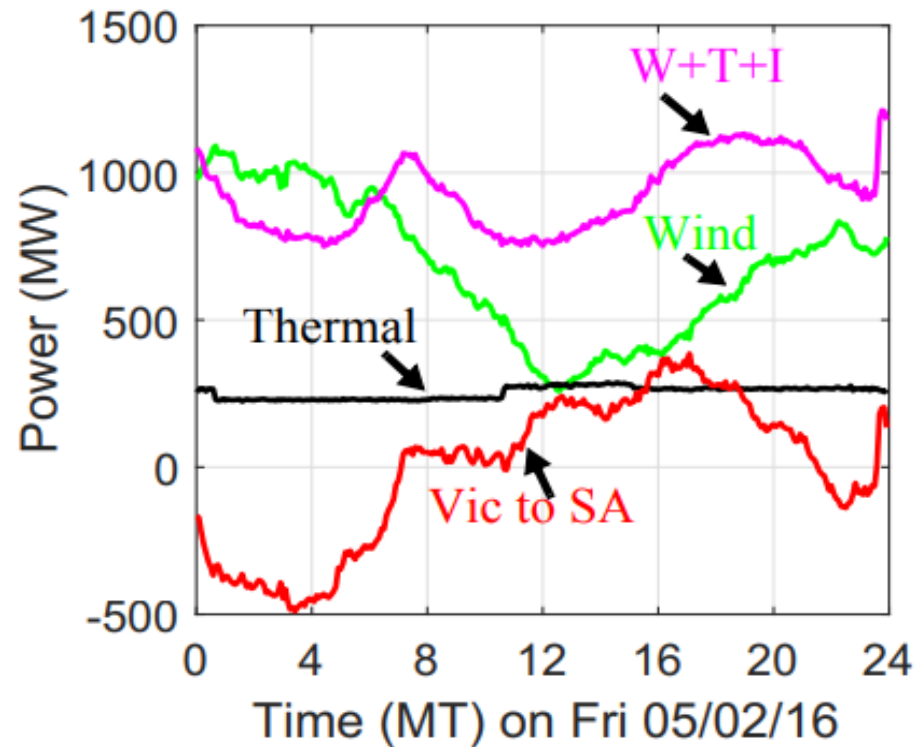
Storage
(Pumped hydro
/ batteries)

Motivation

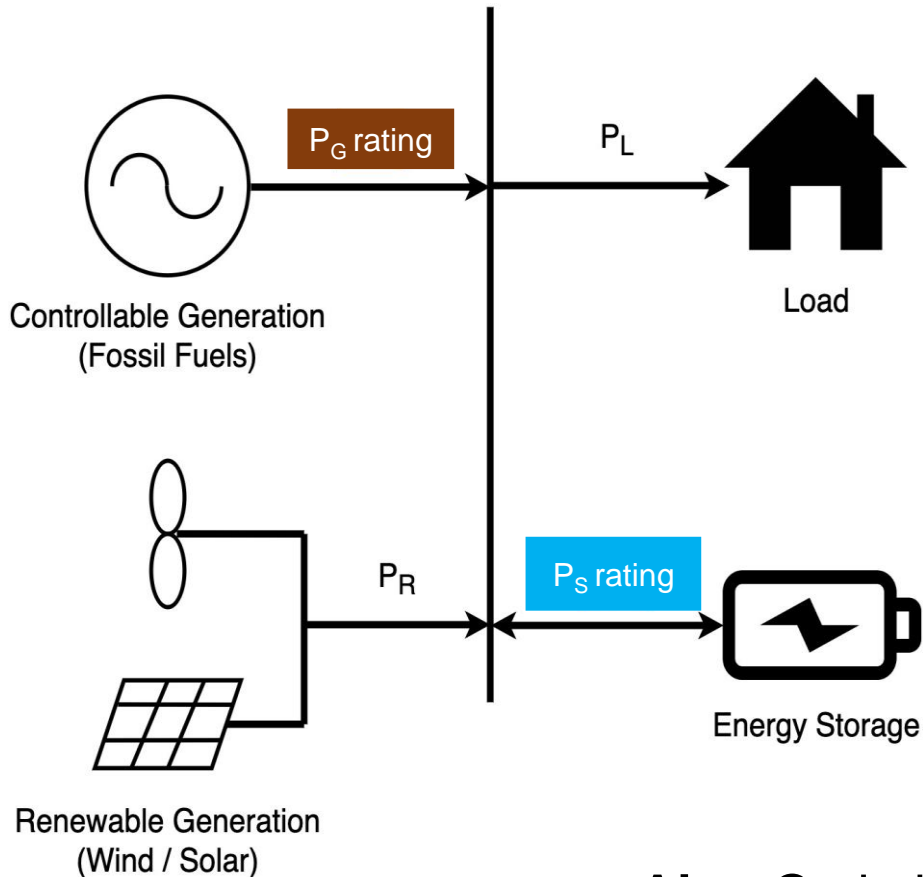
- **Issue:** Intermittent Renewable

Power Sources

- Power may not be available when needed
- Leading to dispatchability issue
 - inability to control power supply
- **Solution:** Energy Storage



How to Optimise?

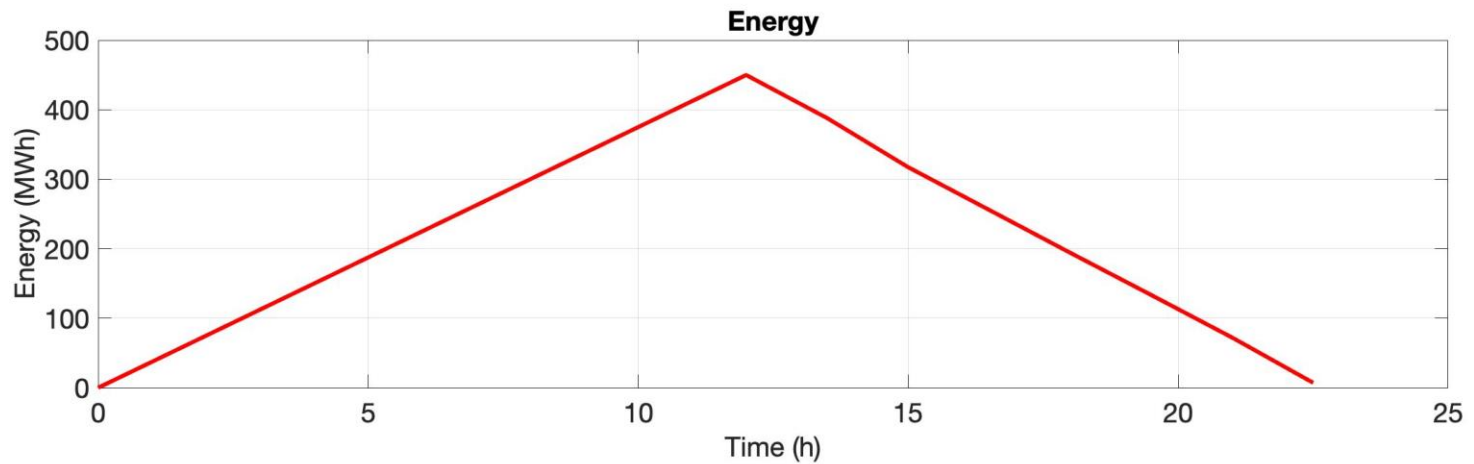
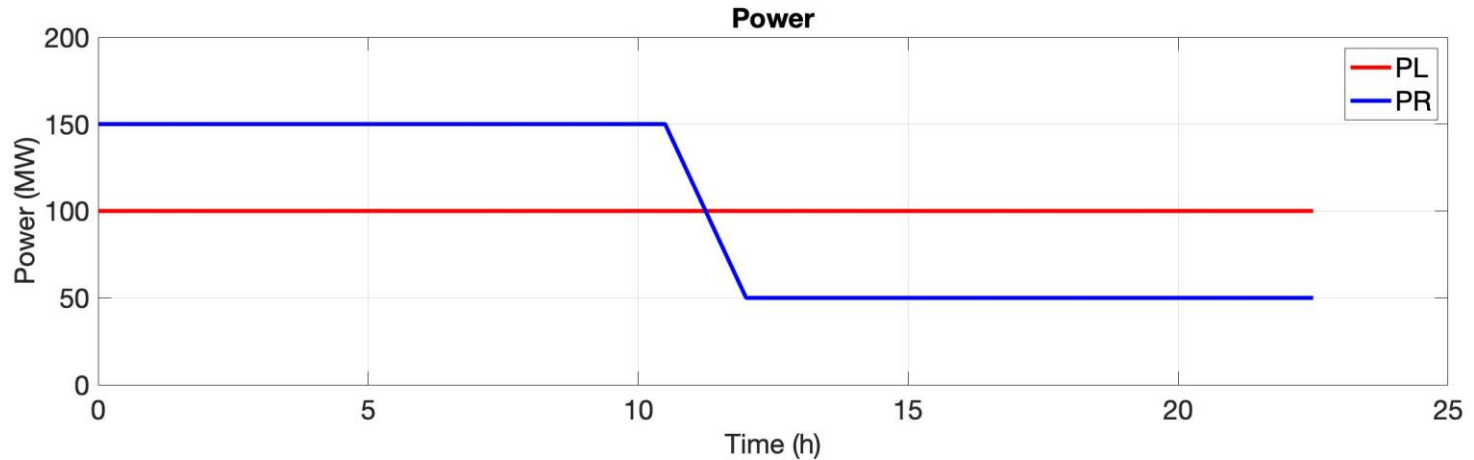


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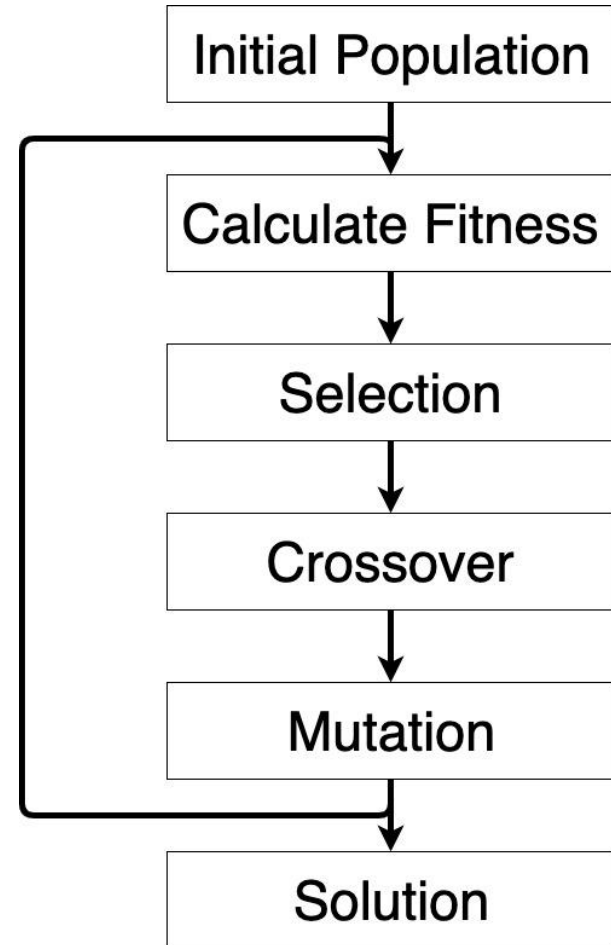
Aim: Optimise P_S rating to minimise P_G rating

Why Optimise?



Genetic Algorithms

- Search based algorithm.
- Use the concept of “Survival of the Fittest”
- Find an optimal solution through a set of processes.
- Repeatedly modifies population of individual solutions.



Why Genetic Algorithms?

Number of time variables (t) \rightarrow
(48 variables)

	$v_{1,1}$	$v_{1,2}$	\dots	$v_{1,t}$
	$v_{2,1}$	$v_{2,2}$	\dots	$v_{2,t}$
	\vdots	\vdots	\ddots	\vdots
	$v_{Npop,1}$	$v_{Npop,1}$	\dots	$v_{Npop,t}$

(100 individuals)
Number of individuals ($Npop$) \rightarrow

Technique 1: Code Vectorisation

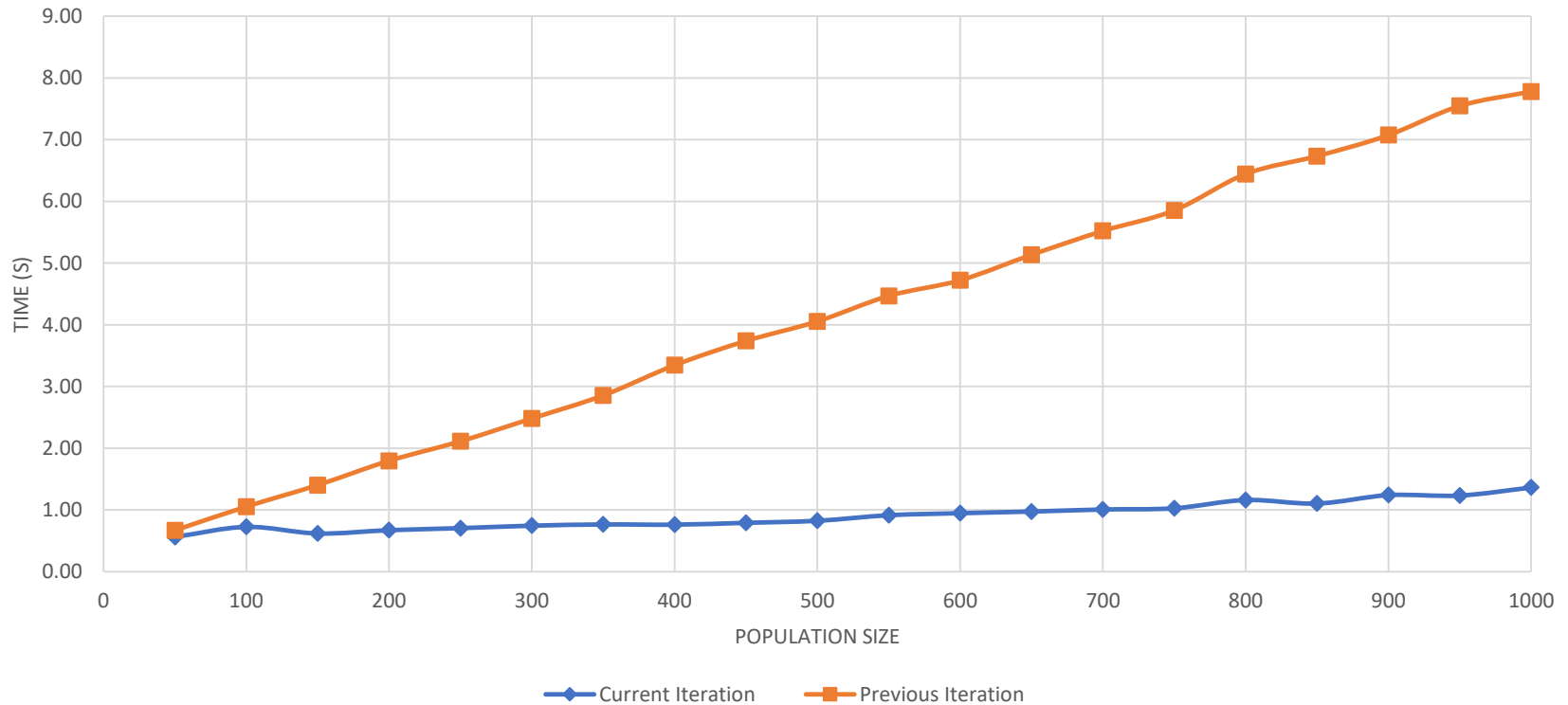
- Serialised Code → Vectorised Code
 - To increase speed

Serialised Code	Vectorized Code
<pre>for i = 1:10 y(2, i) = randVal; end</pre>	<pre>y(2, all) = randVal;</pre>

- **Serialised Code** : Low speed → Takes more time to process
- **Vectorized Code** : High speed → Takes less time to process

Results: Code Vectorisation

BENCHMARK 1: TIME (S) VS POPULATION SIZE

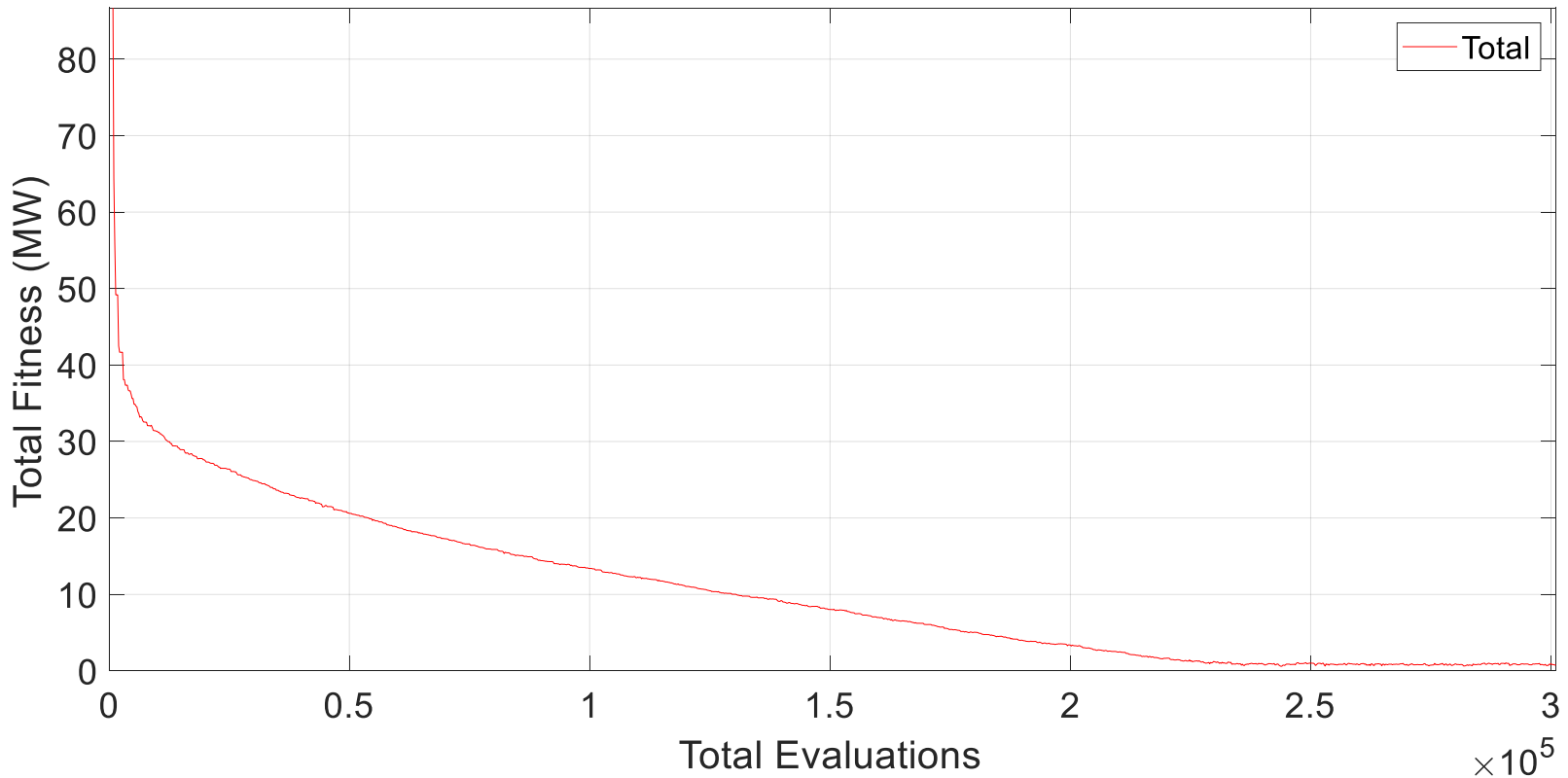


- The current iteration (**blue**) is **5 times faster** than the previous iteration (**orange**)

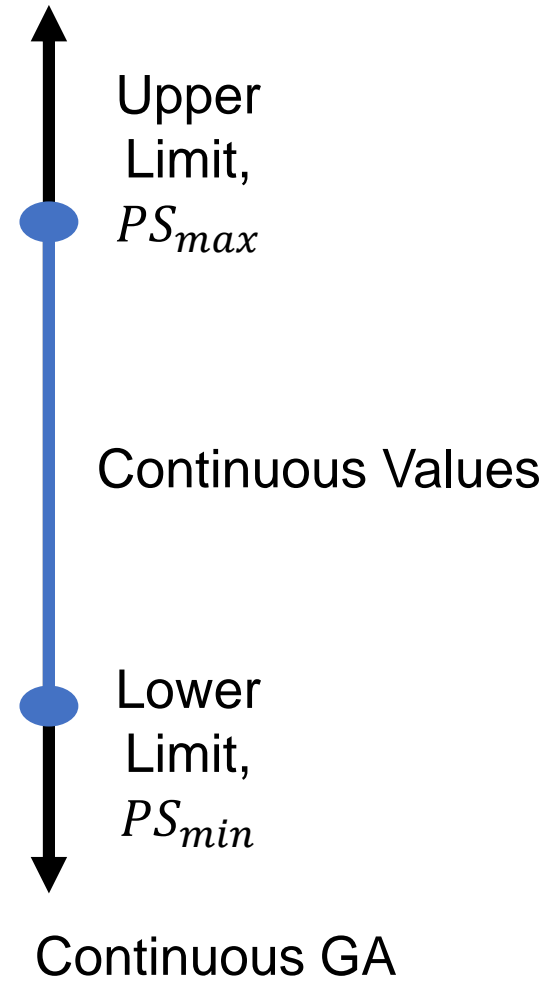
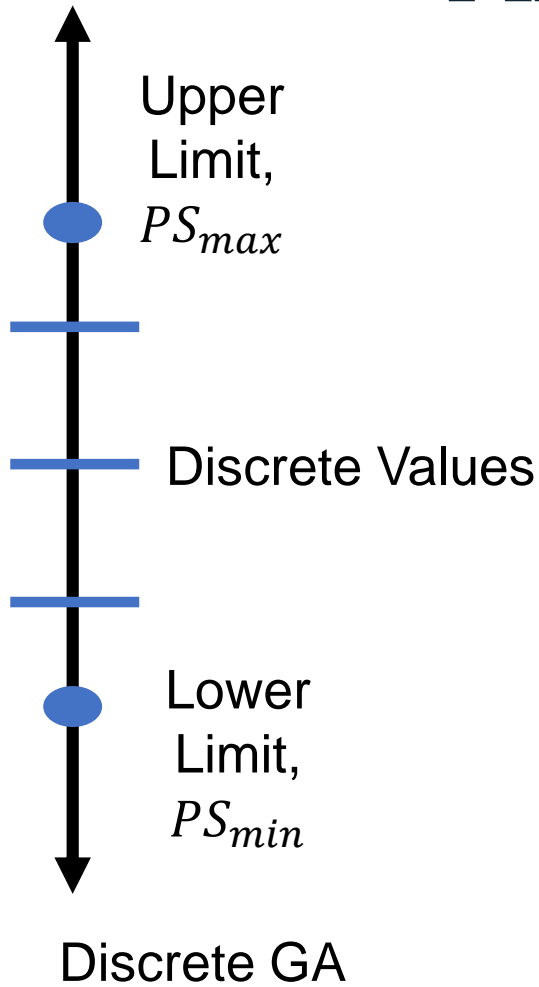
Technique 2: Convergence Theorem

Benchmark case 2: PR = PL = PSET, 48 samples - Nstep=701 (1MW)

Total Fitness



Technique 3: Continuous Genetic Algorithm



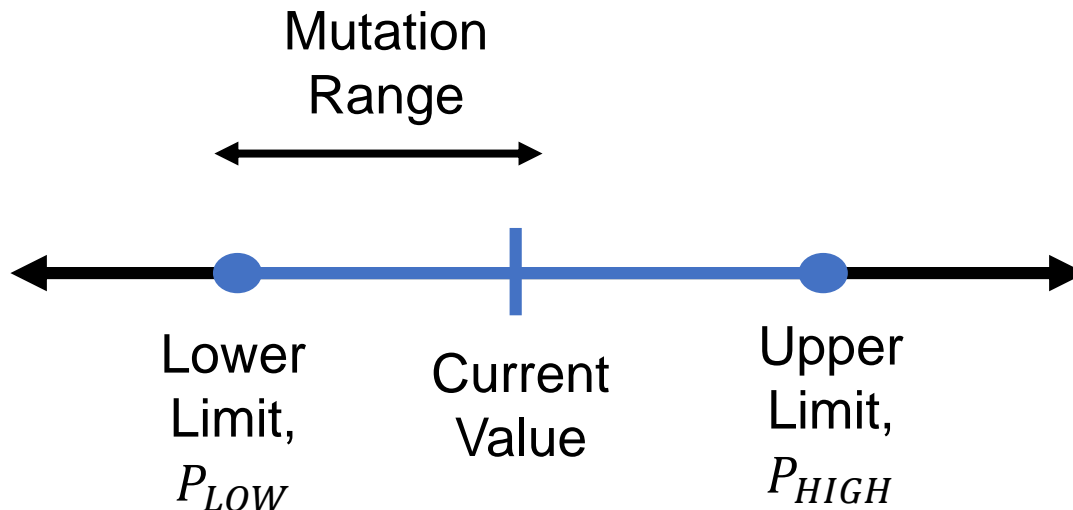
Results: Continuous Genetic Algorithm

Test Case 2	
Continuous Final Fitness (MW)	Discrete Final Fitness (MW)
0.458	2.000

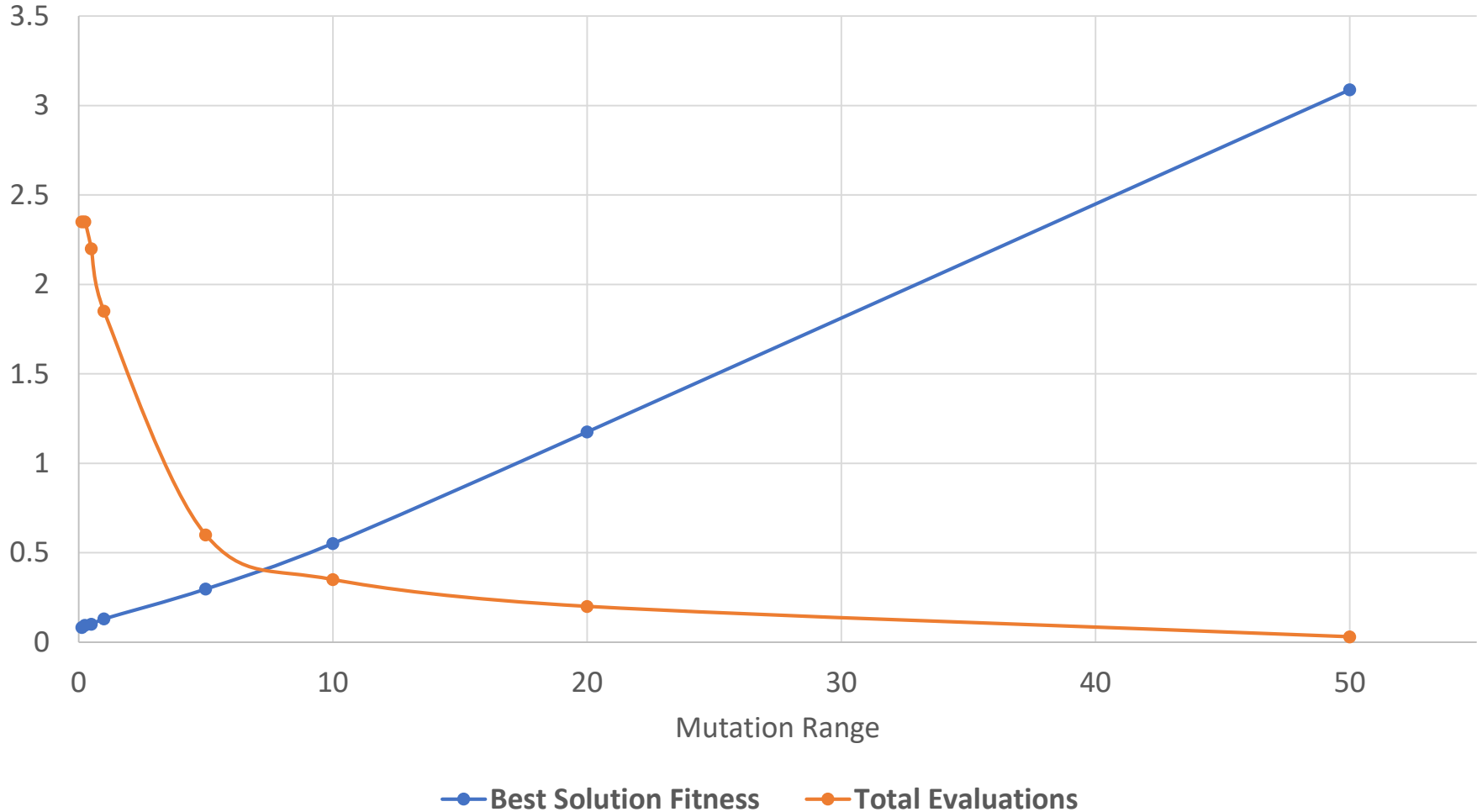
- Continuous GA is more accurate because it has a lower fitness.
- Lower fitness means less fossil fuels.

Technique 4: Mutation Range

- This only applies to Continuous GA
- The range of values that can be mutated to from an initial value
- To reduce the variability of the mutation process and increase performance



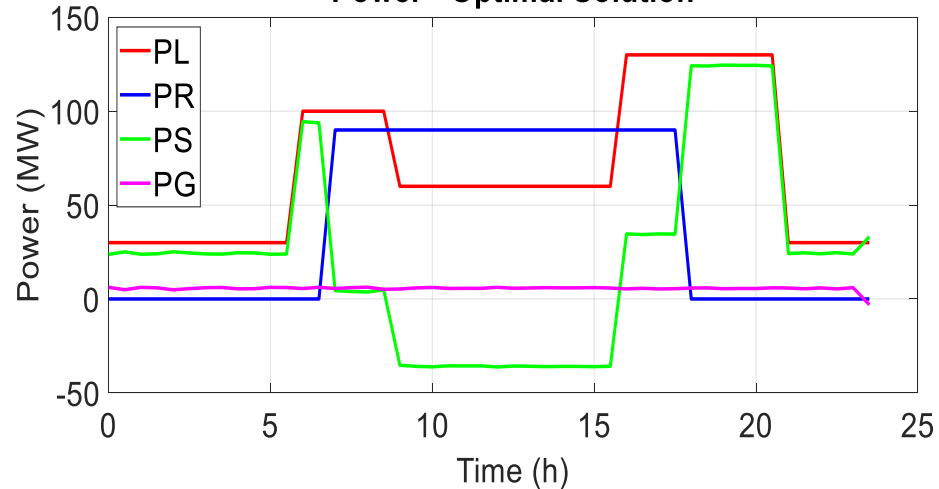
Results: Mutation Range



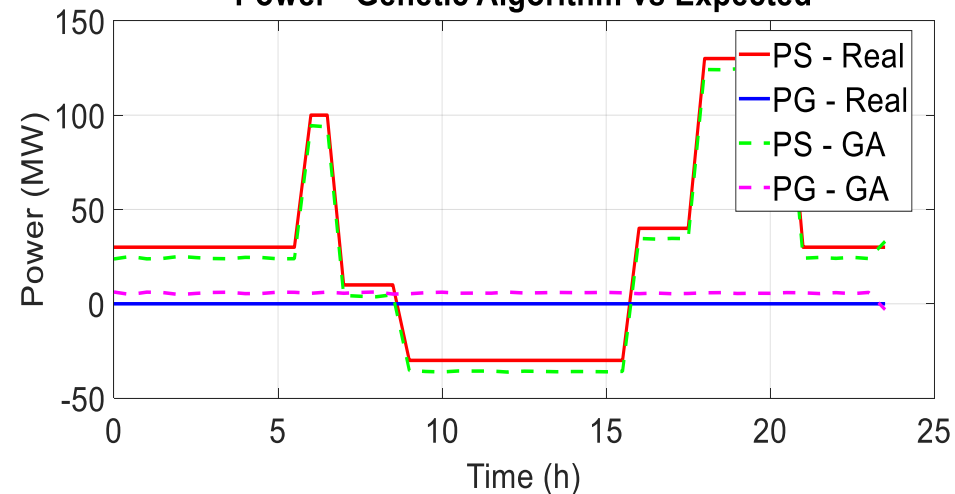
Accuracy Benchmarking

- Solar PV Benchmark
- P_L and P_R varies throughout the day.
- Power Model Results:
 - Deviation in P_S & P_G by 6 MW
 - GA is reasonably accurate here

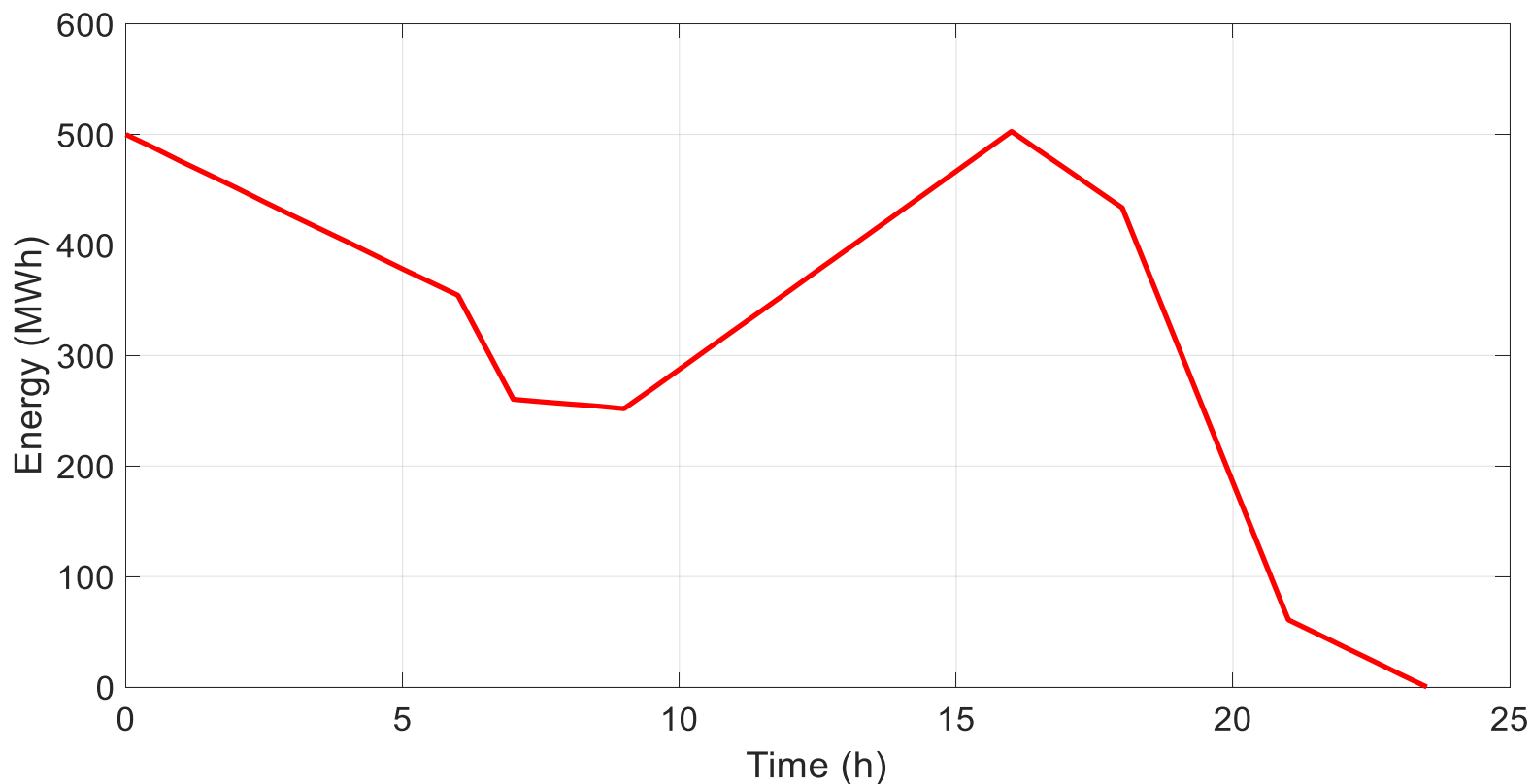
Benchmark case 7: Solar PV Benchmark Case - From EPS
Power - Optimal Solution



Power - Genetic Algorithm vs Expected



Accuracy Benchmarks



- Shows the amount of energy within the storage device
- Shows the discharging and charging rates

Future Work and Conclusion

- Future Work
 - Use real-time data (i.e. AEMO)
 - More accurate modelling of the SA power system
 - MEX File
- Conclusion
 - Genetic Algorithm produces faster & more accurate results.
 - Provides an estimate for storage needs.

References

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Thank you for listening



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