

Machine Learning in Economic Dispatch Problem

- Deep Q Network based Reinforcement Learning Approach

Deep Q Network (DQN) has achieved gameplay learning successfully. If our engineering problem can be treated as a game, can machine also solve it? In the traditional approach, we have found that for many situations in practice we cannot easily find the optimal marginal cost λ of power generation. In this project, machine learned by itself to solve economic dispatch problems!

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Engineering Problem \rightarrow Machine Gaming \rightarrow Learning \rightarrow Playing \rightarrow Engineering Solution

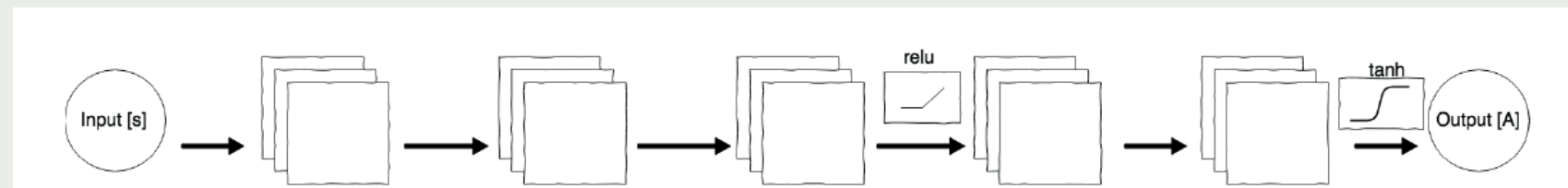
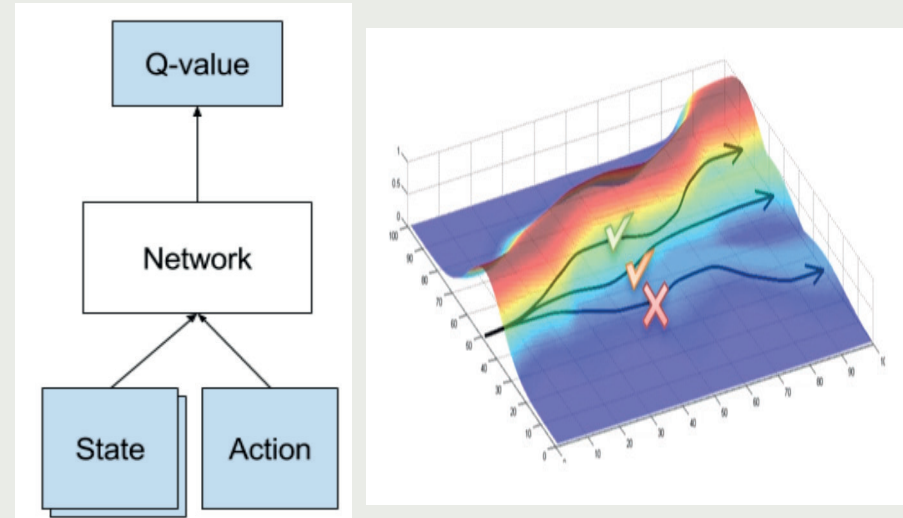


Figure 14: Neural Network for Actor: $A(s) = a; s \in \mathbb{R}^{n_s}, a \in (-1, 1)^{n_a}$

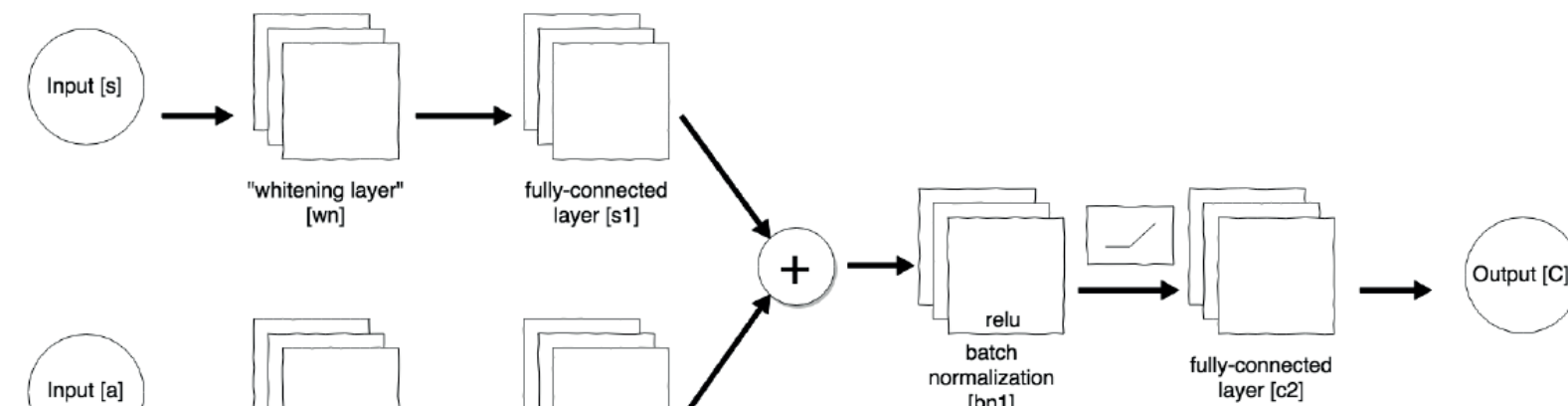
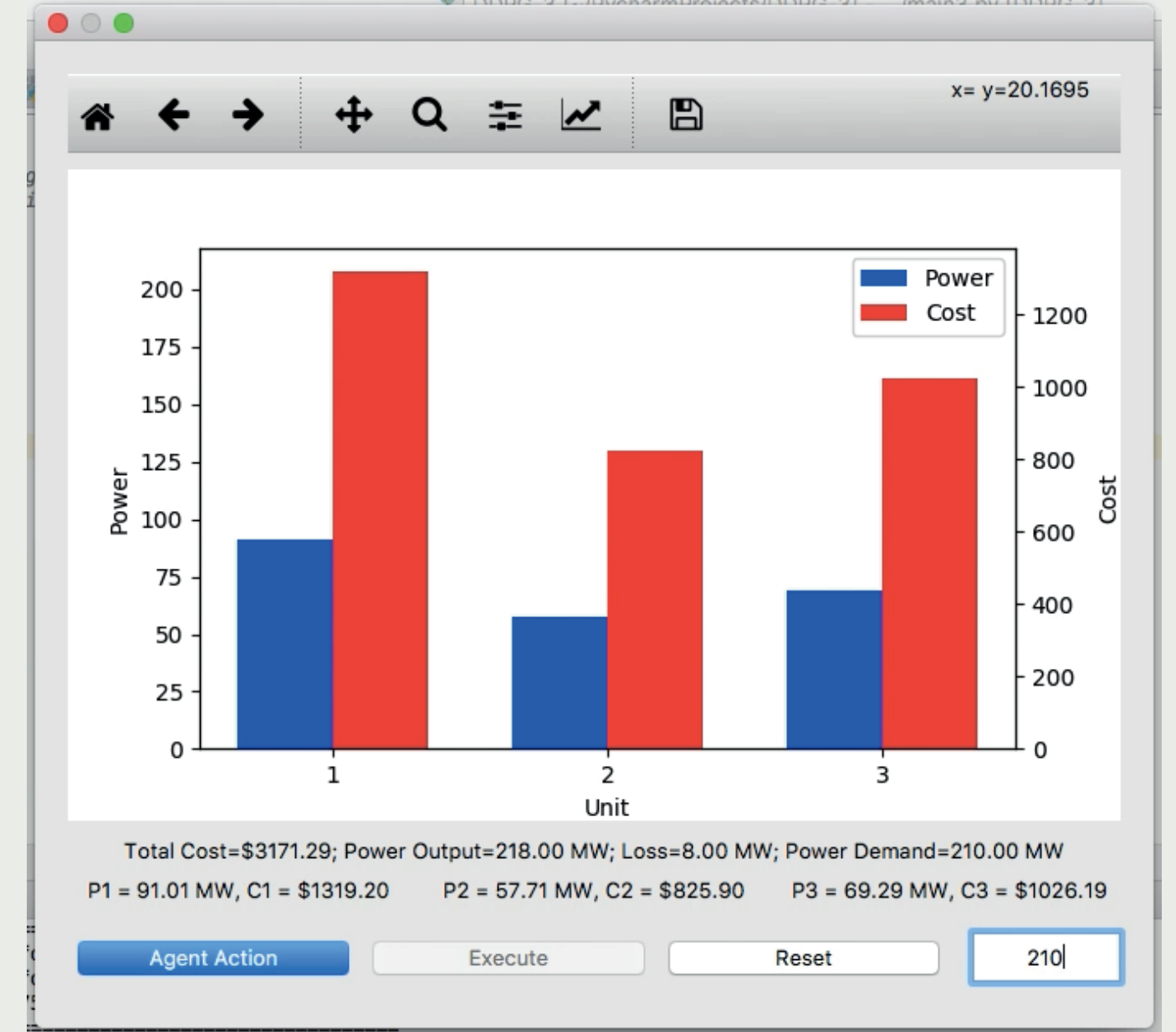


Figure 15: Neural Network for Critic: $C(s, a) = Q; s \in \mathbb{R}^{n_s}, a \in (-1, 1)^{n_a}, Q \in \mathbb{R}^1$



DQN
Actor
Critic
Policy

Reinforcement Learning Theory

Deep Deterministic Policy Gradient
Neural Network Model

Engineering Problem as a Game

Evaluate on 3-generator system with demands 140/240/340/440MW

Train with full-connected hidden layers consisted of 300 neurous

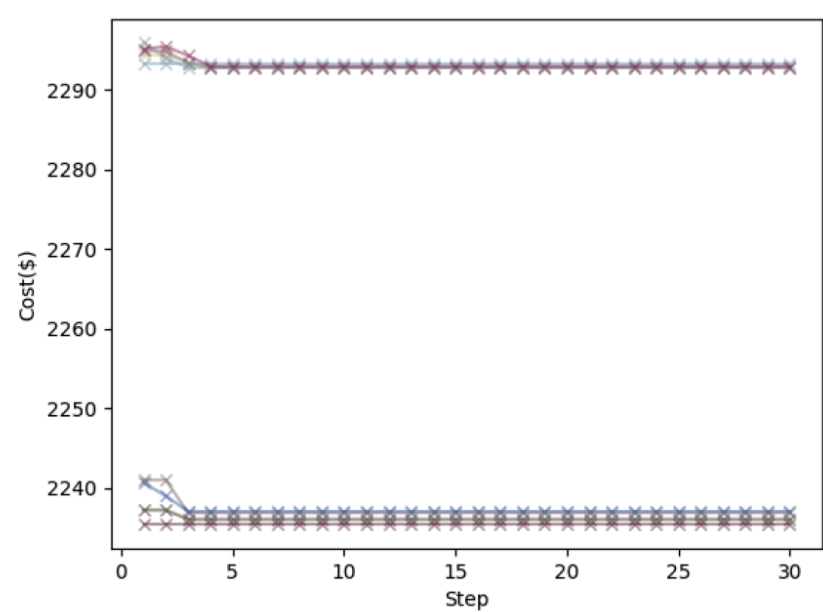


Figure 26: Optimizing the cost for $P_d = 140$ MW (lossy/lossess)

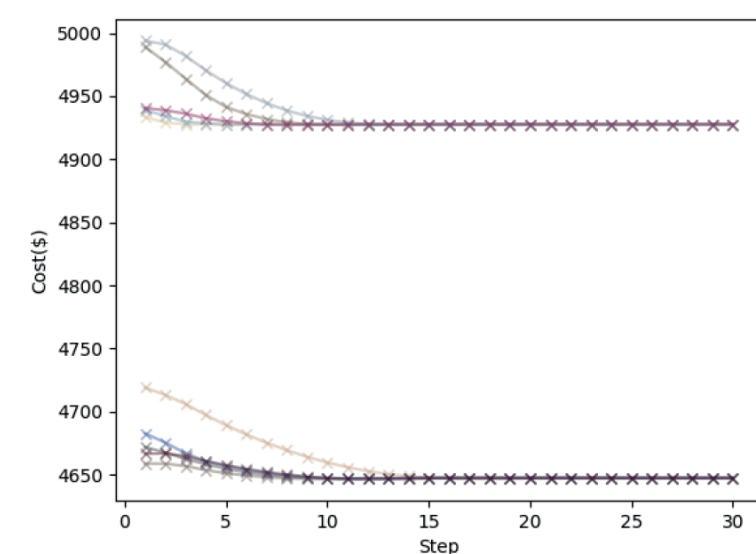


Figure 28: Optimizing the cost for $P_d = 340$ MW (lossy/lossess)

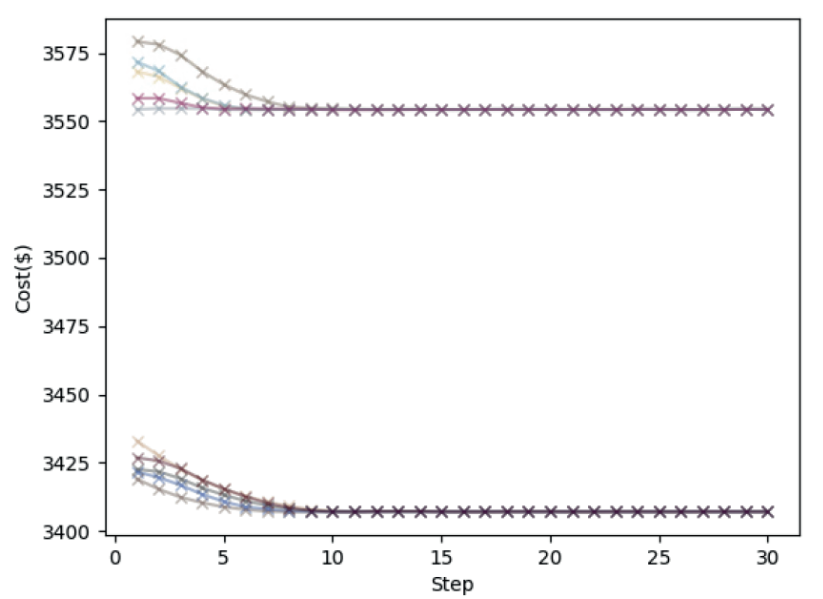


Figure 27: Optimizing the cost for $P_d = 240$ MW (lossy/lossess)

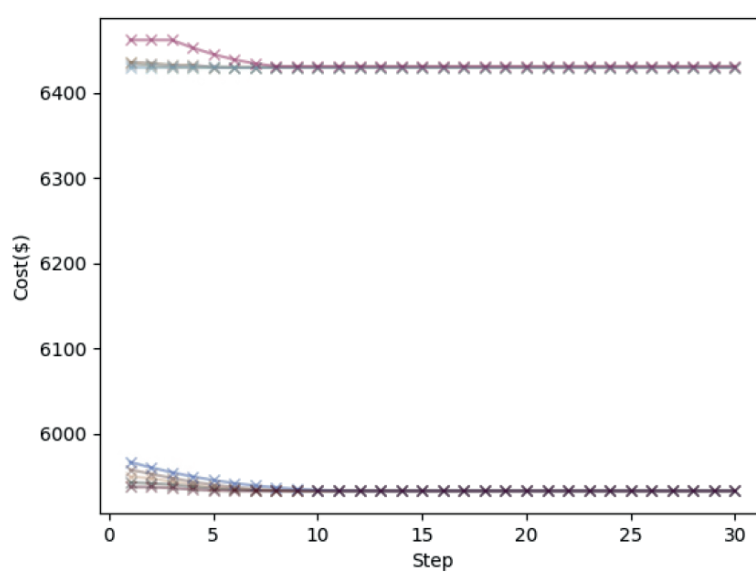
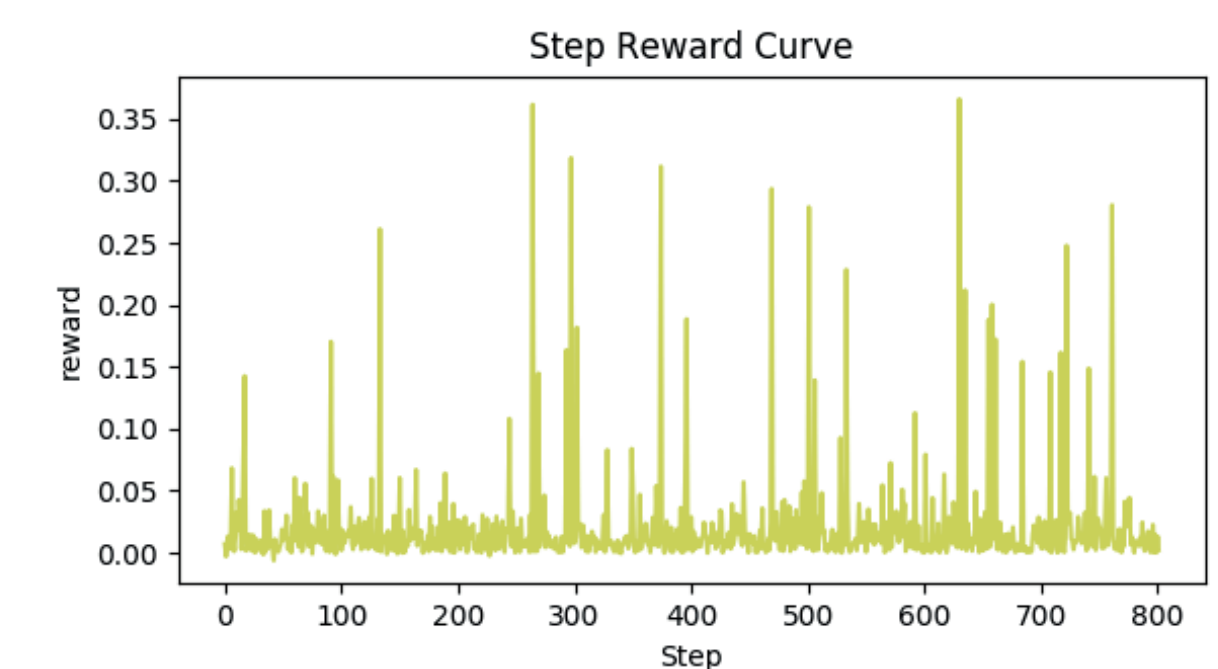
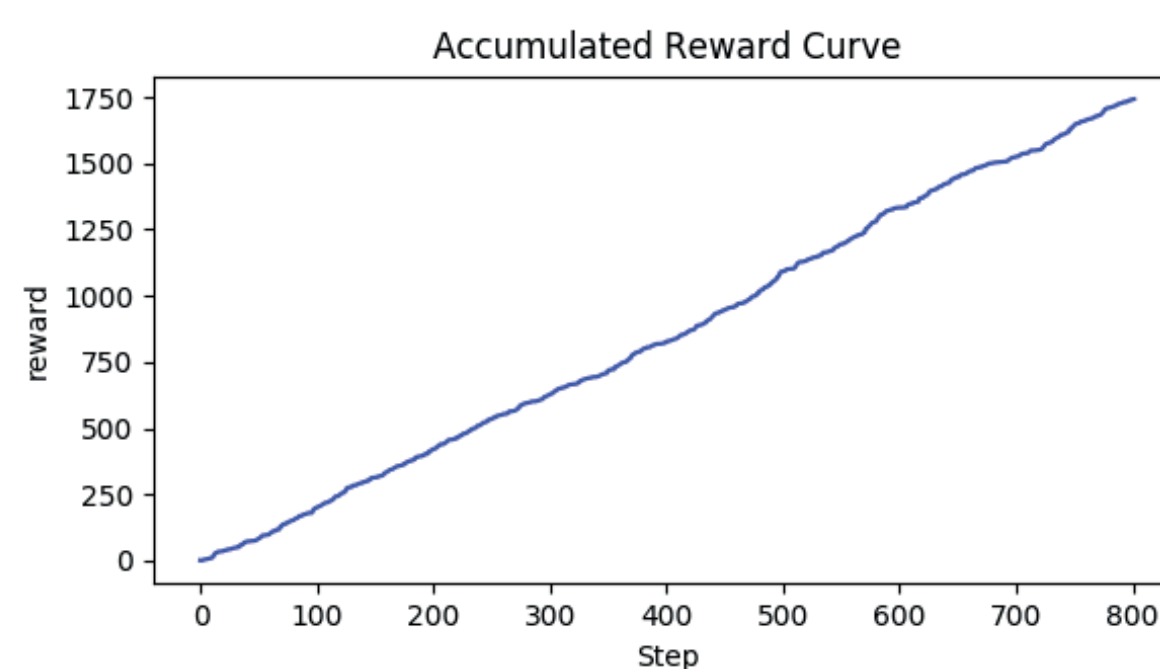
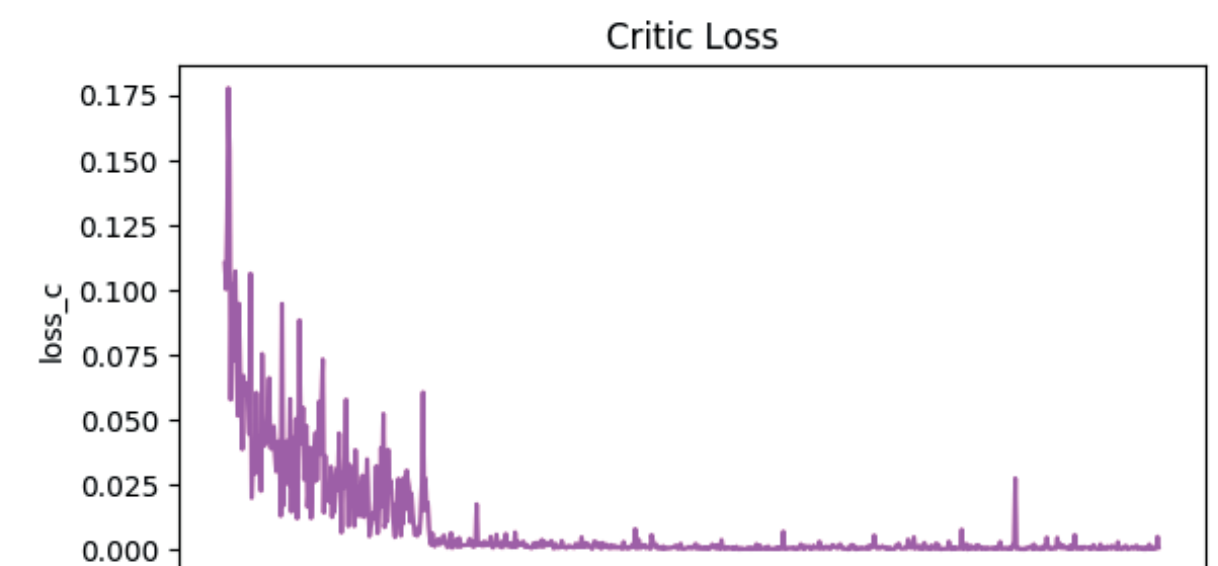
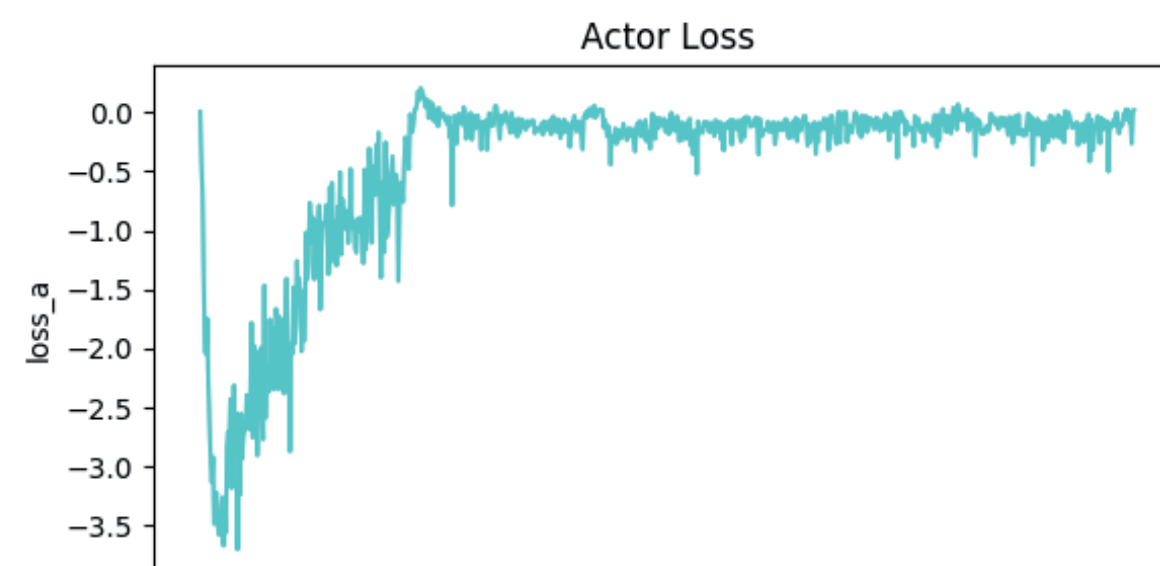
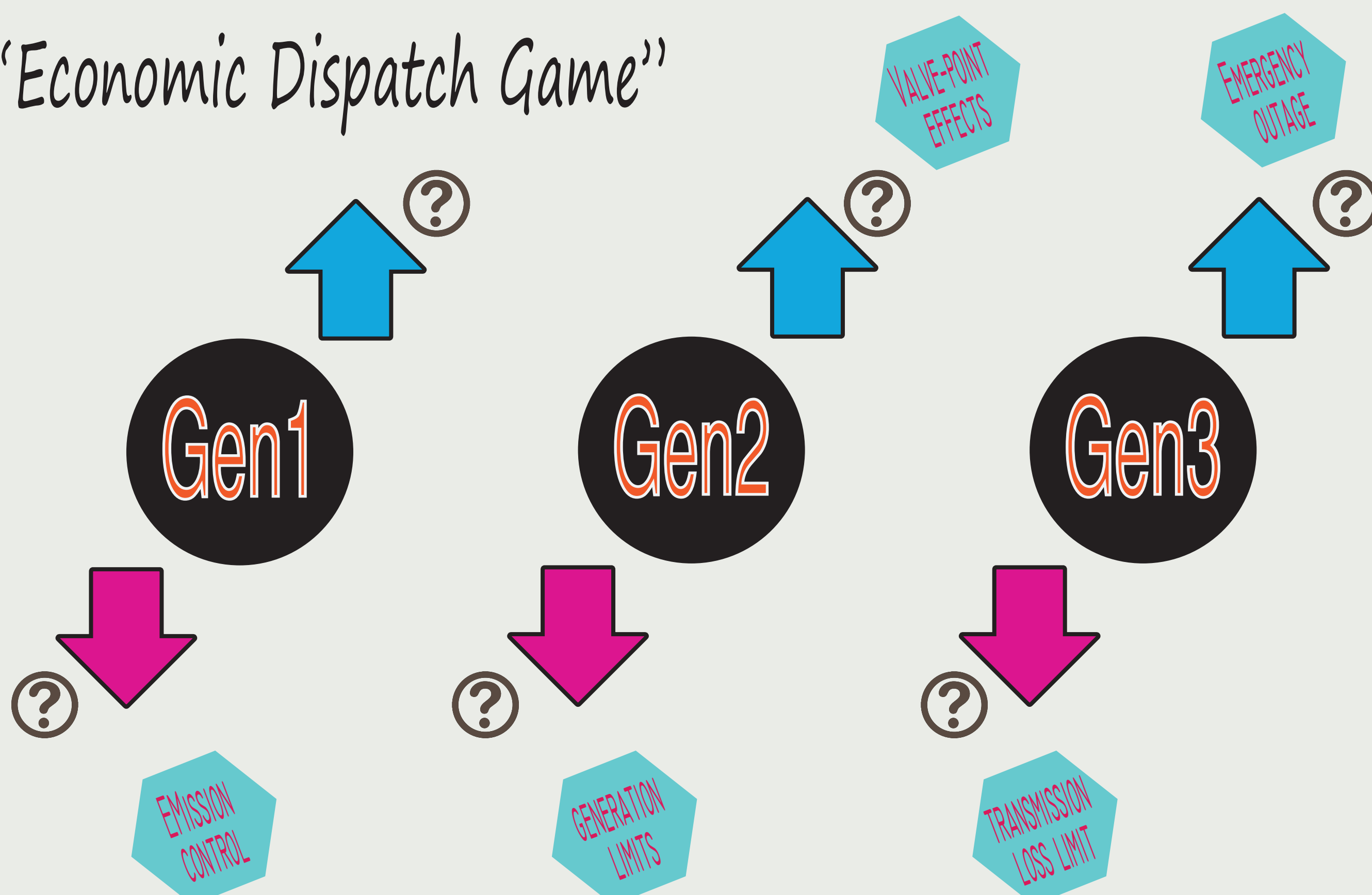


Figure 29: Optimizing the cost for $P_d = 440$ MW (lossy/lossess)



A single training gives all solution on a same case with various scenarios!

“Economic Dispatch Game”



>> Objectives:

- 1) Well-trained network can give us a good solution much faster than the stochastic searching for repetitive usages
- 2) No re-learning is required for the change of designed conditions

>> How?

I made an imaginary game that allows the machine understand the relationship between dispatches and the change of total cost. Neural Network is applied to remember and learn the results.

>> How the game works?

The rewards is the cost reduction for a state transition.

positive \rightarrow good action negative \rightarrow bad action

Theoreticality the agent can pass through all obstacles and find the best way to obtain highest cost reduction from any initial state!

>> It works? Any conclusion?

It works in a small system, which is a good start. However if the system gets complicated, much deeper network is required. For a large and practical environment, the agent may have to take days in learning with a very deep network.