1. Comparison of Execution Time and Convergence

Model-Based (Policy Iteration / Value Iteration):

- Execution Time: These methods can be computationally expensive per iteration because they require sweeping through all states and actions and using the full transition probability model.
- o Convergence: They generally converge faster (fewer iterations) since they exploit the exact environment dynamics (state transition probabilities and rewards).
- o Overall: High per-iteration cost but fewer iterations → efficient when the model is known and small/medium state space.

• Model-Free (Q-Learning):

- Execution Time: Each step is relatively cheap since it only updates one stateaction pair based on sampled experience.
- Convergence: Convergence is slower and depends on exploration strategy, learning rate, and number of episodes. It requires significantly more interactions with the environment to approximate the optimal policy.
- o Overall: Low per-step cost but requires many more steps → efficient in large/unknown environments but slower to converge.

2. Difference Between Model-Based and Model-Free Algorithms

Model-Based:

- The agent has or learns a model of the environment (transition probabilities + rewards).
- Uses this model to plan ahead (simulate outcomes) and derive optimal policies.
- o Example: Value Iteration, Policy Iteration.
- o Strength: Faster convergence when the model is accurate.
- Weakness: Requires complete knowledge of the environment, which is often unrealistic.

Model-Free:

- The agent does not use an explicit model of the environment.
- o Learns policies or value functions directly from experience through trial and error.
- o Example: Q-Learning, SARSA.
- o Strength: Works even when the environment's dynamics are unknown.
- Weakness: Slower convergence, depends heavily on exploration.



