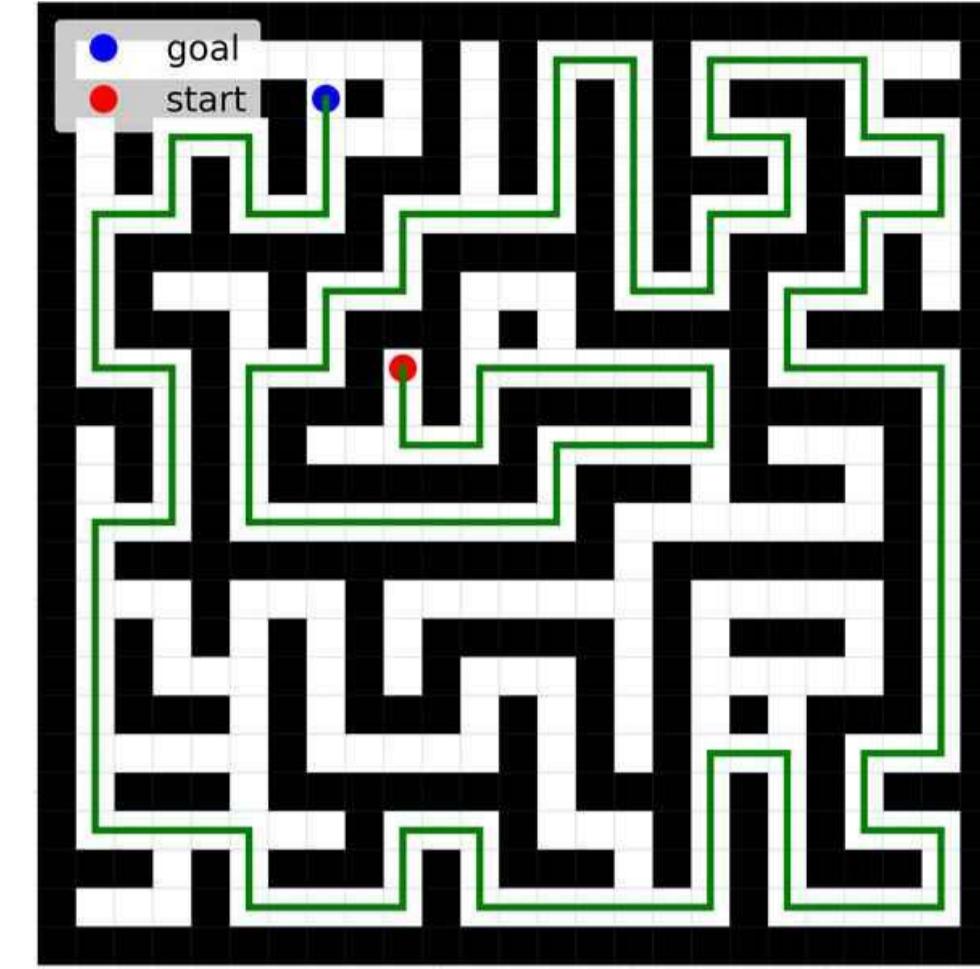


Highway Value Iteration Networks

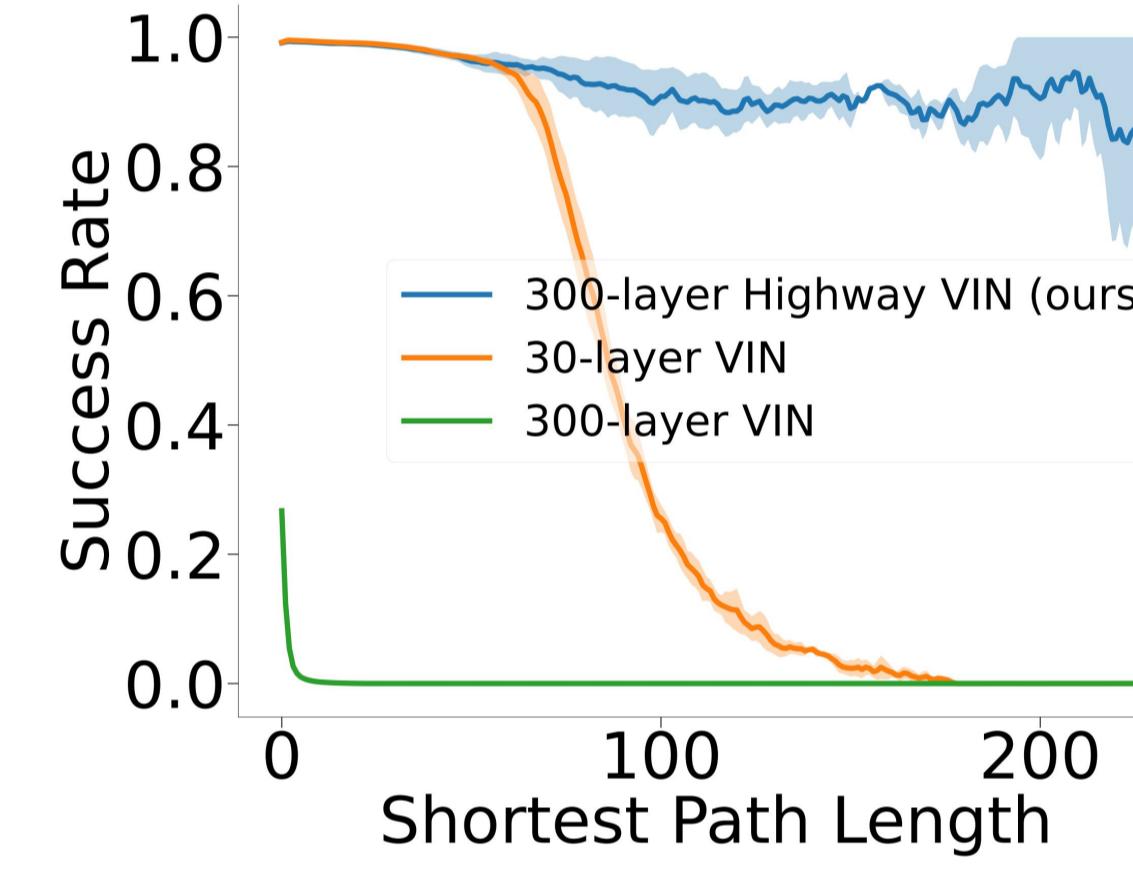
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Motivation

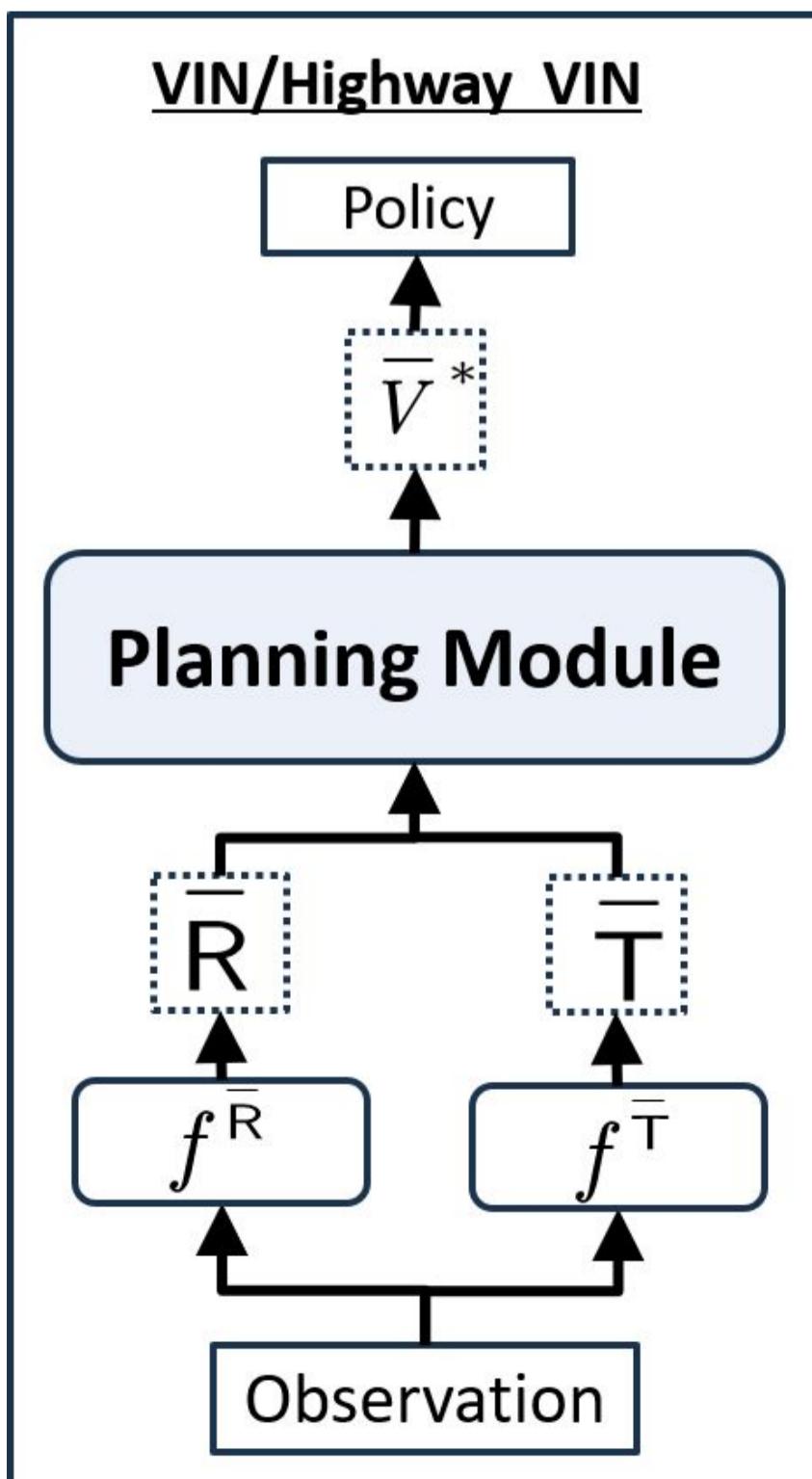


Maze Navigation



Performance on tasks with various shortest path lengths

Background



Architecture of Value Iteration Network (VIN) [1]

Method

Bellman Optimality/Expectation Operator

$$(\mathcal{B}V)(s) \triangleq \max_a \sum_{s'} \mathcal{T}(s'|s, a) [\mathcal{R}(s, a, s') + \gamma V(s')], \quad (\mathcal{B}^\pi V)(s) \triangleq \sum_a \pi(a|s) \sum_{s'} \mathcal{T}(s'|s, a) [\mathcal{R}(s, a, s') + \gamma V(s')]$$

Planning Module of VIN

$$V^{(n+1)} = \mathcal{B}V^{(n)}$$

Planning Module of VIN

$$V^{(n)} \rightarrow \text{Value Iteration Module}$$

Value Iteration Module

$$\max_{\bar{Q}} \bar{Q}(\cdot)$$

Value Iteration Module

$$\text{conv}_{2 \times F \times F}(\cdot; \bar{T})$$

Value Iteration Module

$$\bar{V}^{(0)}$$

Value Iteration Module:

$$(1) \bar{Q}_{\bar{a}, i, j}^{(n)} = \sum_{i', j'} \left(\bar{T}_{\bar{a}, i', j'} \bar{R}_{i-i', j-j'} + \bar{T}_{\bar{a}, i', j'} \bar{V}_{i-i', j-j'}^{(n-1)} \right)$$

$$(2) \bar{V}_{i, j}^{(n)} = \max_{\bar{a}} \bar{Q}_{\bar{a}, i, j}^{(n)}$$

Value Exploration Module:

$$(1) \bar{Q}_{\bar{\pi}, \bar{a}, i, j}^{(n+n_b)} = \sum_{i', j'} \left(\bar{T}_{\bar{a}, i', j'} \bar{R}_{i-i', j-j'} + \bar{T}_{\bar{a}, i', j'} \bar{V}_{\bar{\pi}, i-i', j-j'}^{(n+n_b-1)} \right)$$

$$(2) \bar{V}_{n_p, i, j}^{(n+n_b)} = \sum_{\bar{a}} \bar{\pi}_{n_p, \bar{a}, i, j}^{(n+n_b)} \bar{Q}_{n_p, \bar{a}, i, j}^{(n+n_b)}$$

Aggregate Gate:

$$\bar{V}_{i, j}^{(n+N_b)} = \sum_{n_p=1}^{N_p} \tilde{A}_{n_p, i, j}^{(n+N_b)} \sum_{n_b=1}^{N_b} A_{n_p, i, j}^{(n+N_b)} \max_{\bar{a}} \left\{ \bar{V}_{n_p, i, j}^{(n+n_b)}, \bar{V}_{i, j}^{(n+1)} \right\}$$

where

$$\tilde{A}_{n_p, i, j}^{(n+N_b)} = \frac{\exp(\alpha_A \bar{V}_{n_p, i, j}^{(n+N_b)})}{\sum_{n_p} \exp(\alpha_A \bar{V}_{n_p, i, j}^{(n+N_b)})} \quad A_{n_p, i, j}^{(n+N_b)} = \frac{\exp(\alpha_A \bar{V}_{n_p, i, j}^{(n+N_b)})}{\sum_{n_b} \exp(\alpha_A \bar{V}_{n_p, i, j}^{(n+N_b)})}$$

Planning Module of Highway VIN

Planning Module of Highway VIN

$$V^{(n+1)} = \operatorname{smax}_{\pi \in \Pi}^{\alpha} \operatorname{smax}_{n \in N}^{\alpha} \max \left\{ (\mathcal{B}^\pi)^{(N_b-1)} \mathcal{B}V^{(n)}, \mathcal{B}V^{(n)} \right\}$$

Planning Module of Highway VIN

$$\bar{V}^{(n)} \rightarrow \text{Highway Block}$$

Value Exploration Module

$$\text{linear}_{|\bar{A}| \times 1 \times 1}(\cdot; \bar{\pi}_{n_p})$$

Value Iteration Module

$$\text{conv}_{2 \times F \times F}(\cdot; \bar{T})$$

Value Iteration Module

$$\bar{V}_{n_p}^{(0)}$$

Value Iteration Module:

$$(1) \bar{Q}_{\bar{a}, i, j}^{(n)} = \sum_{i', j'} \left(\bar{T}_{\bar{a}, i', j'} \bar{R}_{i-i', j-j'} + \bar{T}_{\bar{a}, i', j'} \bar{V}_{i-i', j-j'}^{(n-1)} \right)$$

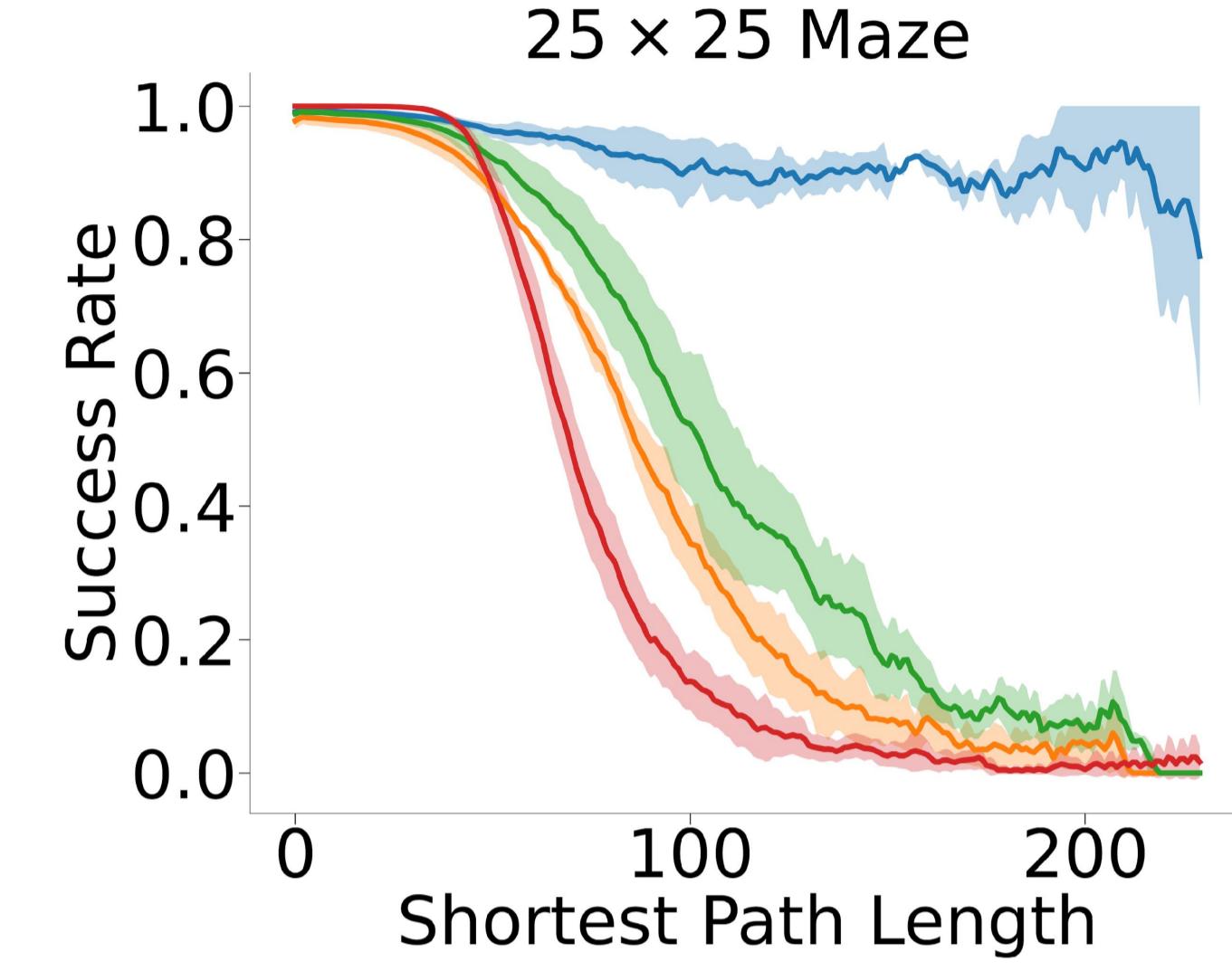
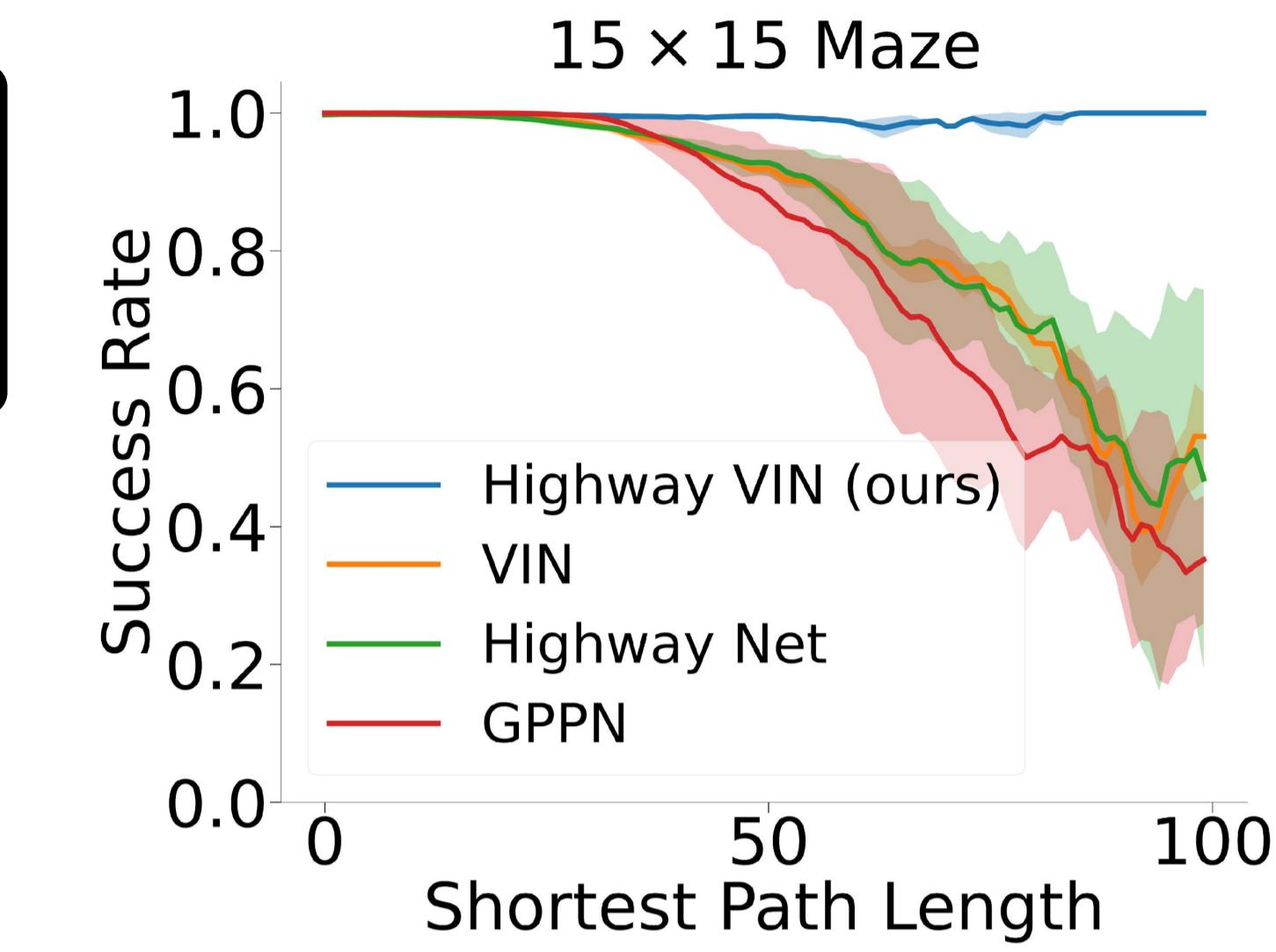
$$(2) \bar{V}_{i, j}^{(n)} = \max_{\bar{a}} \bar{Q}_{\bar{a}, i, j}^{(n)}$$

Value Exploration Module:

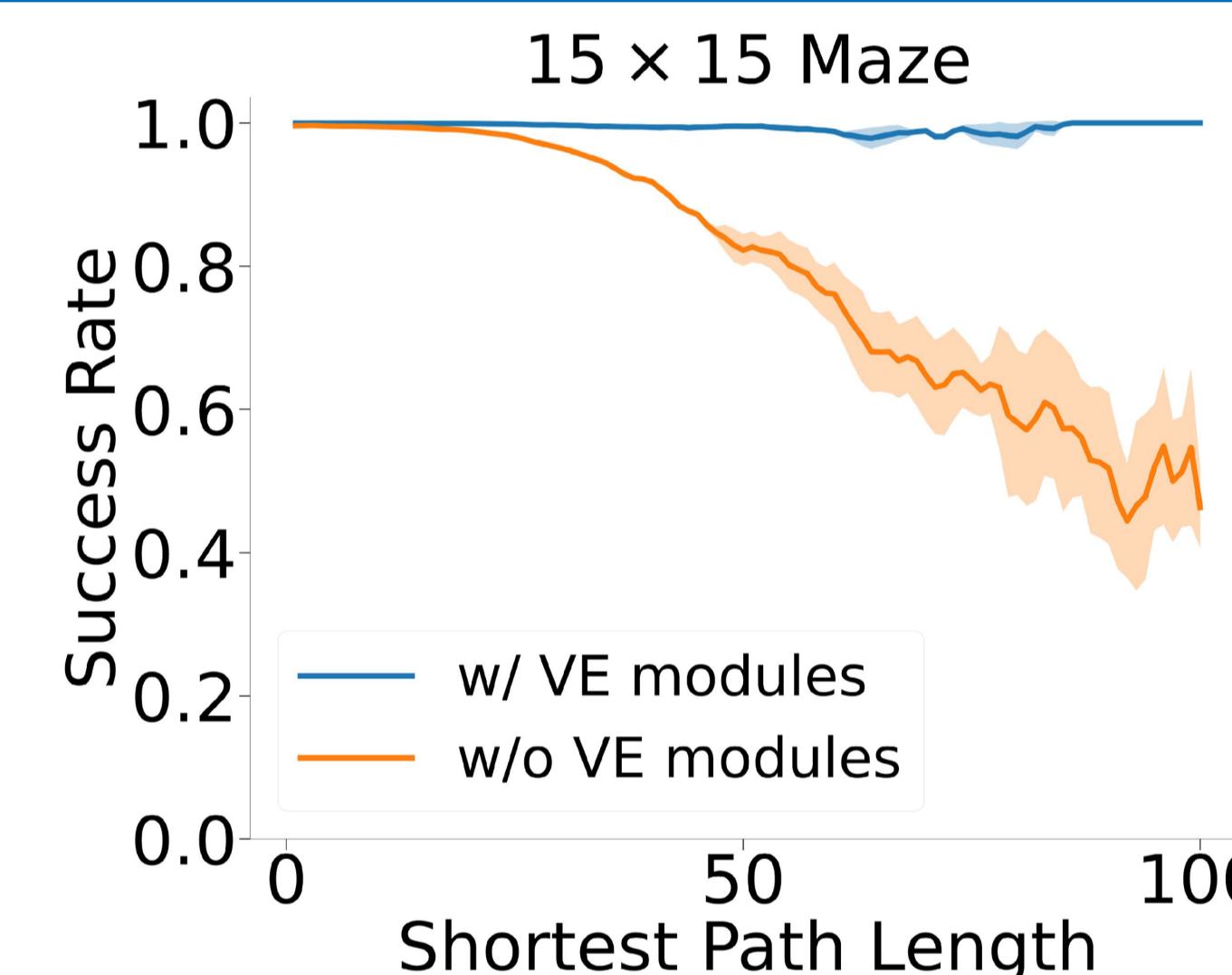
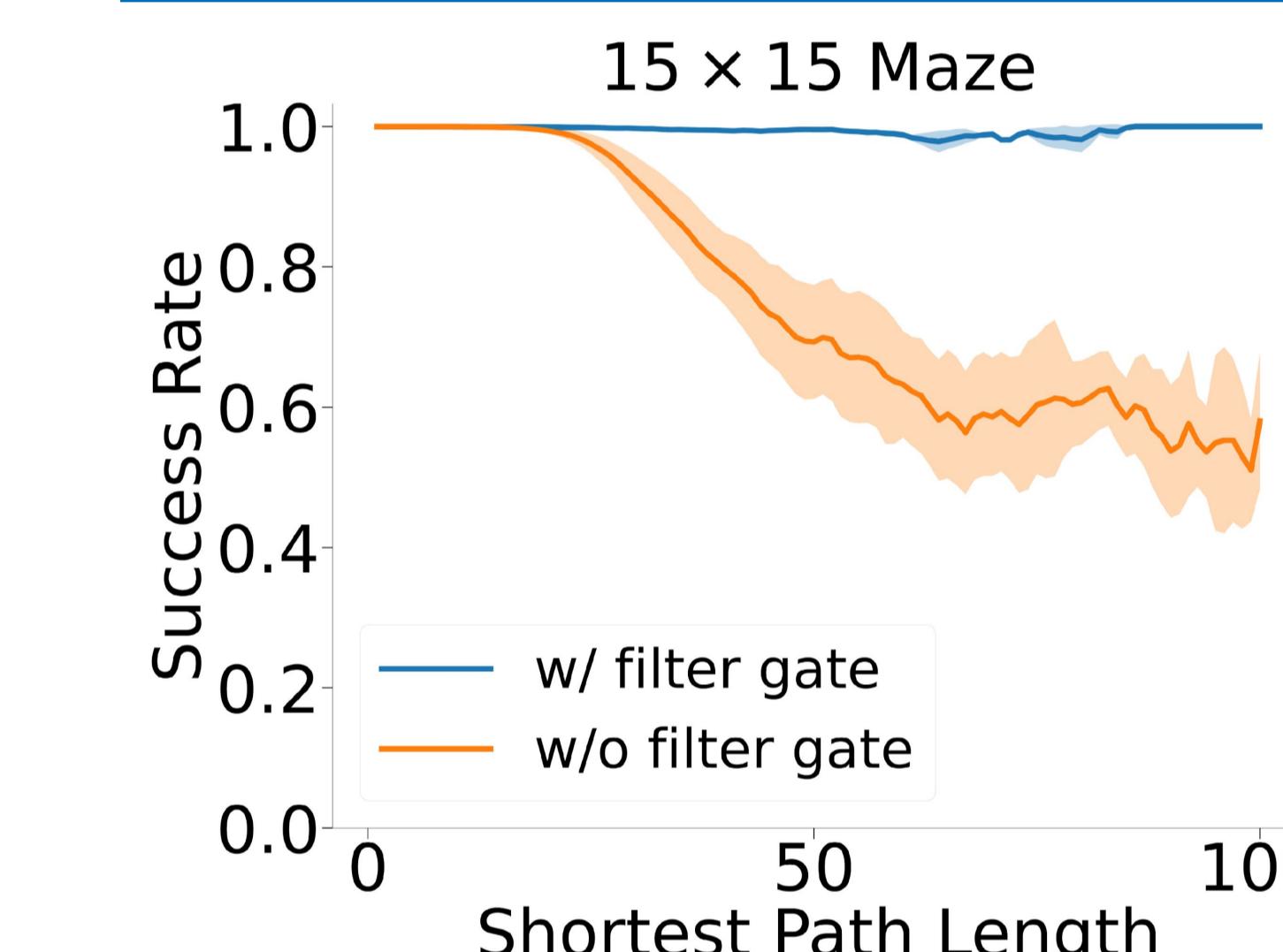
$$(1) \bar{Q}_{\bar{\pi}, \bar{a}, i, j}^{(n+n_b)} = \sum_{i', j'} \left(\bar{T}_{\bar{a}, i', j'} \bar{R}_{i-i', j-j'} + \bar{T}_{\bar{a}, i', j'} \bar{V}_{\bar{\pi}, i-i', j-j'}^{(n+n_b-1)} \right)$$

$$(2) \bar{V}_{n_p, i, j}^{(n+n_b)} = \sum_{\bar{a}} \bar{\pi}_{n_p, \bar{a}, i, j}^{(n+n_b)} \bar{Q}_{n_p, \bar{a}, i, j}^{(n+n_b)}$$

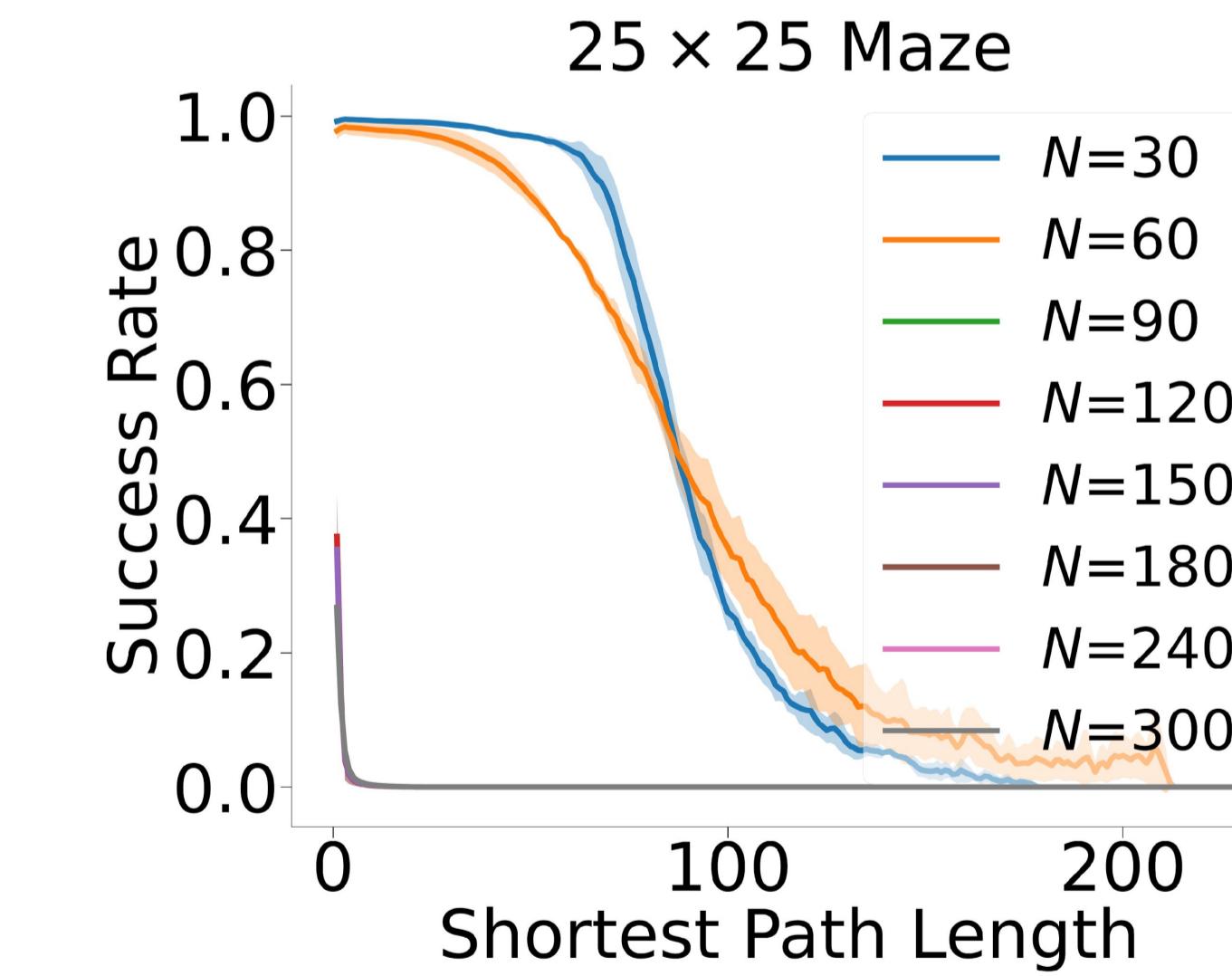
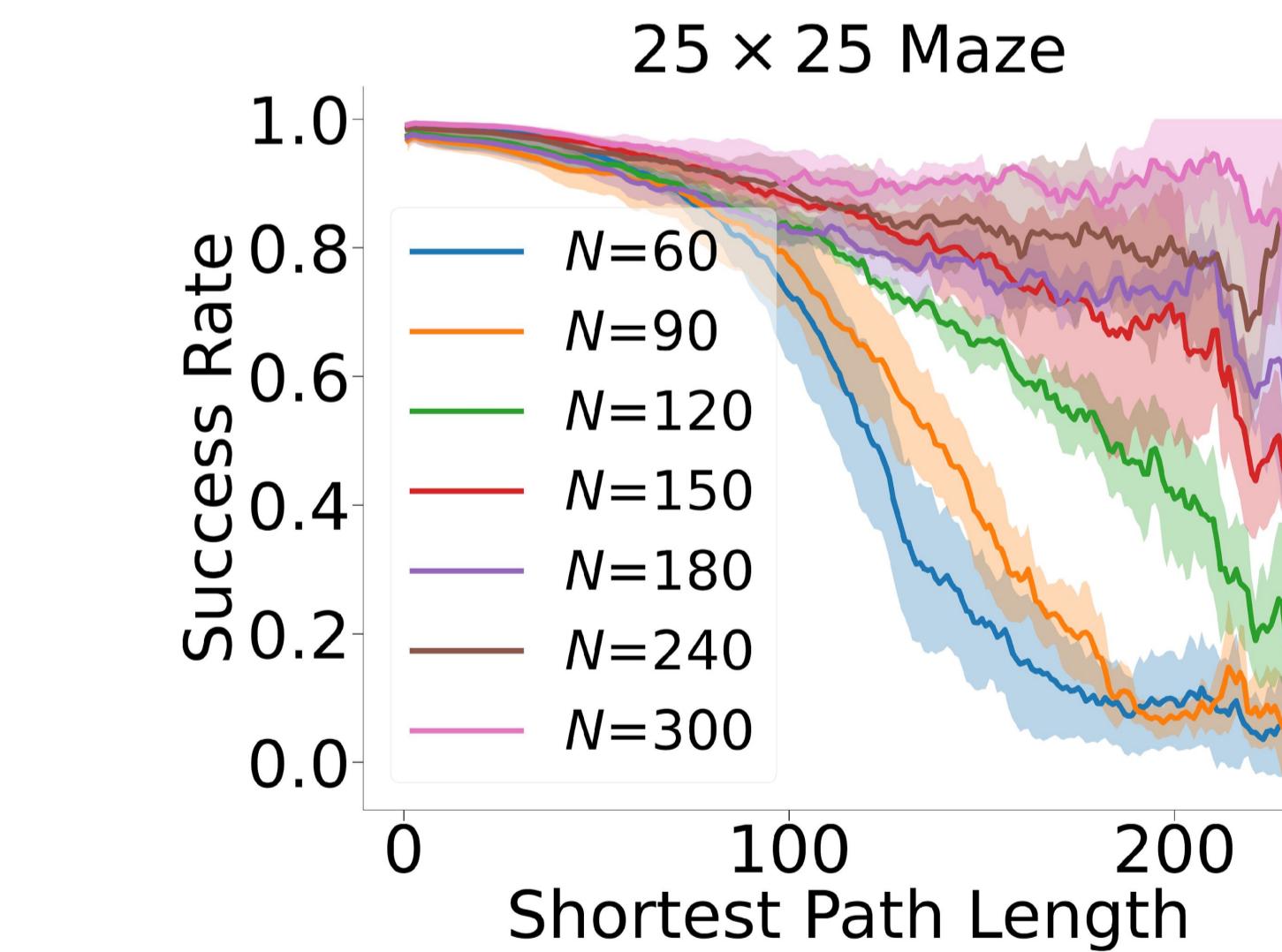
Experiments



Ablation Studies



Filter Gate



[1] Tamar A, Wu Y, Thomas G, Levine S, & Abbeel P. Value iteration networks[J]. Advances in neural information processing systems, 2016, 29.

[2] Wang Y, Strupl M, Faccio F, Wu Q, Liu H, Grudzien M, Tan X, and Schmidhuber J. Highway reinforcement learning[J]. arXiv preprint arXiv:2405.18289, 2024.