

Wide Replacement Products Meet Gray Codes: Toward Optimal Small-Bias Sets

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CCC 2026

Small-bias sets

A set $S \subseteq \{0,1\}^n$ is ε -biased if

$$\forall 0 \neq \tau \in \{0,1\}^n \quad |\mathbb{E}_{s \sim S} (-1)^{\langle \tau, s \rangle}| \leq \varepsilon$$

- * PRGs against $\text{GF}(2)$ -linear tests.
- * Many applications in complexity (mostly via Fourier analysis).
- * Equivalent to spectral expansion of Cayley graphs over the Boolean cube.
- * Yield binary codes in the high-distance regime.

How small can we get?

- * A random $S \subseteq \{0,1\}^n$ of size $\Theta(n/\varepsilon^2)$ is ε -biased w.h.p.
- * Matches the lower bound up to a $\log(1/\varepsilon)$ factor [MRRW 1977].

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Codes

An ε -biased set $S \subseteq \{0,1\}^n$ of size m yields a binary code with rate $\rho = n/m$ and relative distance $1/2 - \varepsilon$.

Codes corresponding to set size $\Theta(n/\varepsilon^2)$ meet the Gilbert-Varshamov bound.

Explicit constructions

[Naor-Naor STOC 1990]

$$\frac{n}{\varepsilon^{O(1)}}$$

[Alon-Goldreich-Hastad-Peralta FOCS 1990]

$$\frac{n^2}{\varepsilon^2}$$

[BenAroya-TaShma FOCS 2009]

$$\left(\frac{n}{\varepsilon^2}\right)^{5/4}$$

[TaShma STOC 2017]

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Other recent approaches

[Kopparty-TaShma-Yakirevitch CCC 2026]

Algebraic-
Geometric

[C-Doron-Goldgraber-Manket ICALP 2026]

Algebraic-
Geometric

[Hsieh-Mohanty-Zhang FOCS 2026]

Expanders

[Blanc-Doron CCC 2022]

HDX

[TaShma STOC 2017]

$$\frac{n}{\varepsilon^{2+\alpha}}$$

where

$$\alpha = \tilde{O}\left(\left(\frac{1}{\log 1/\varepsilon}\right)^{1/3}\right)$$

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Our quantitative result $1/3 \Rightarrow 1/2$

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Our more significant contribution

Overcoming one of the two barriers toward the holy grail.

Bias Reduction

Observation. If S is an ε_0 -biased set then

$$S + S = \{s_1 + s_2 : s_1, s_2 \in S\}$$

is ε_0^2 -biased.

However, this construction has size $|S|^2 > n^2$.

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Idea [Rozenman-Wigderson]. Derandomize using an expander:

$$S +_G S = \{s_1 + s_2 : s_1 \sim_G s_2\}$$

Claim. This is an $\varepsilon_0^2 + \omega$ -biased set.

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Proof.

A given test τ induces a ± 1 labeling on the vertices of G : vertex s is labeled by $(-1)^{\langle s, \tau \rangle}$. Encode these labels as the diagonal of Π . Then,

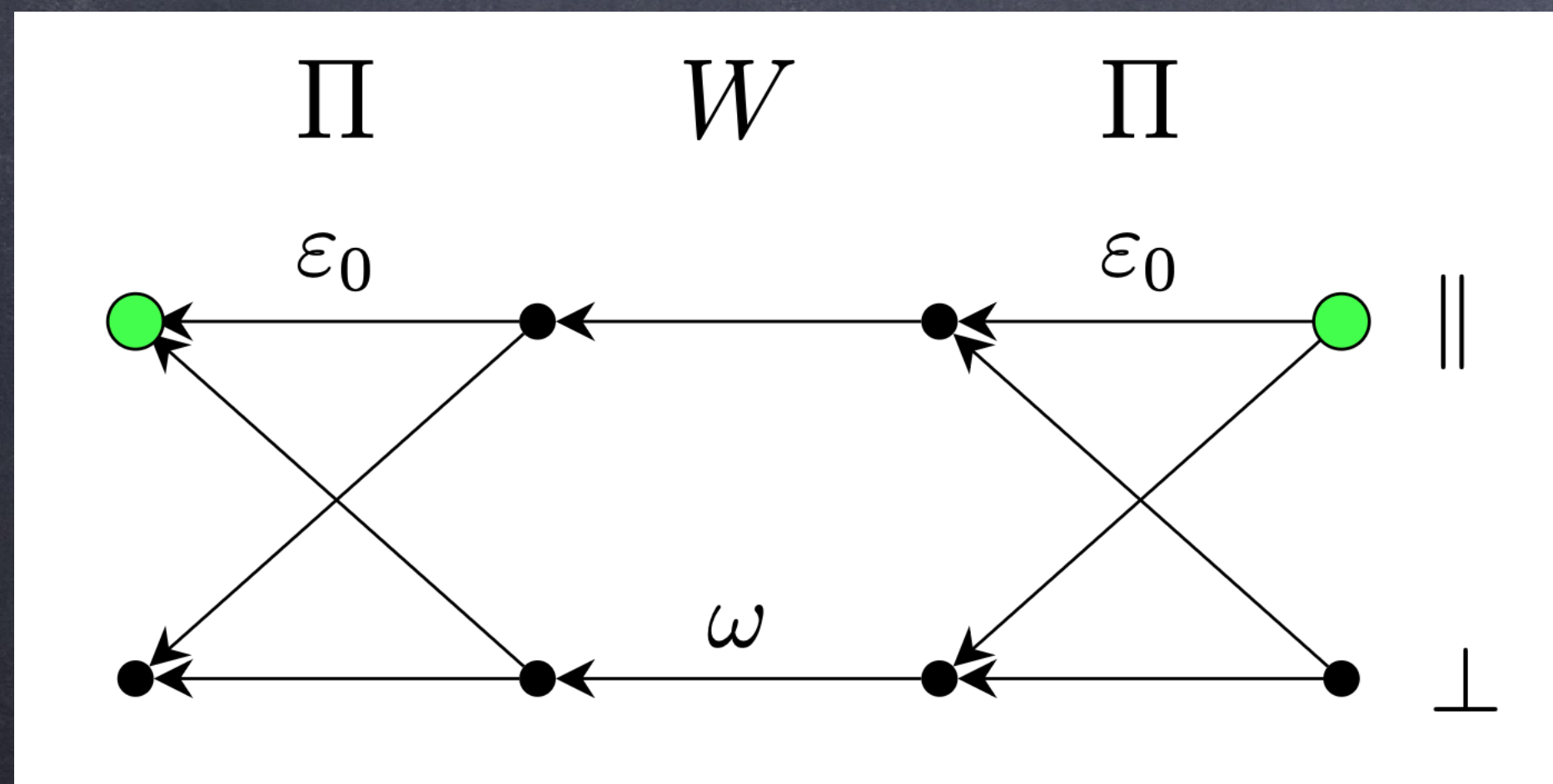
$$\varepsilon_\tau = \left| \mathbf{1}^T \Pi W \Pi \mathbf{1} \right|$$

Claim. $S +_G S = \{s_1 + s_2 : s_1 \sim_G s_2\}$ is an $\varepsilon_0^2 + \omega$ -biased set.

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$$\varepsilon_\tau \leq \varepsilon_0^2 + \omega$$

Longer walks - dream version

2-vertex walks yield bias $\varepsilon_0^2 + \omega$.

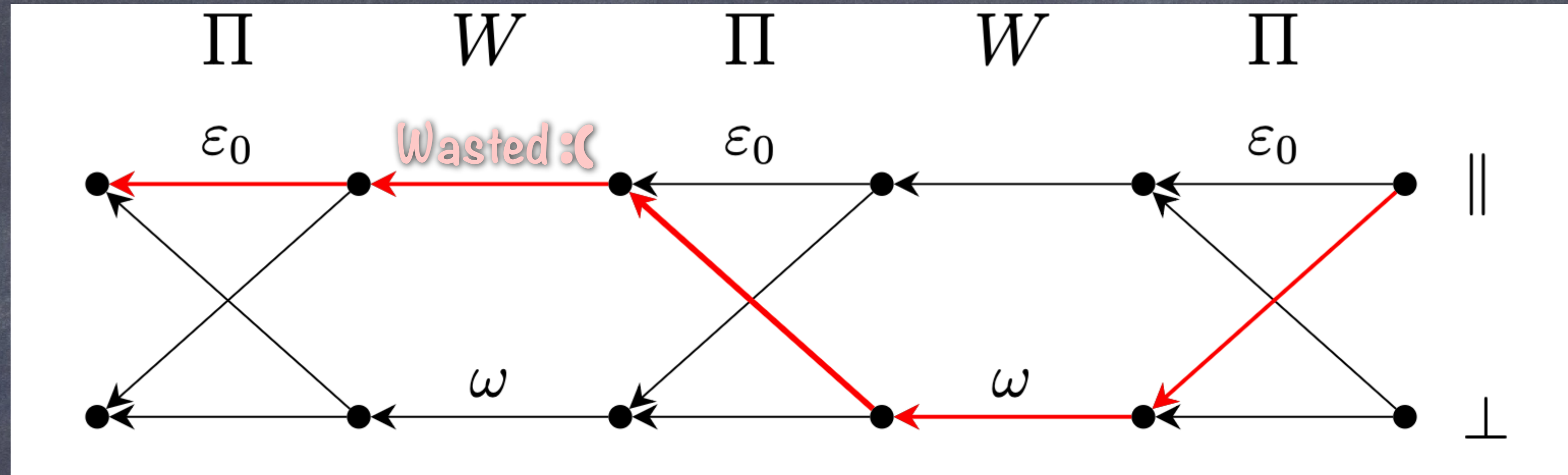
If t -vertex walks give bias $\varepsilon_0^t + \omega^t$ (or $\omega^{t-O(1)}$) then, using a Ramanujan graph for G , starting with $\varepsilon_0 \approx O(1)$, and taking $t = O(\log 1/\varepsilon)$ yields an ε -bias set of size

$$n \cdot \tilde{O}(\varepsilon^{-2})$$

The factor 2 comes from the $\sqrt{}$ in the Ramanujan bound $\omega \approx 2/\sqrt{d}$.

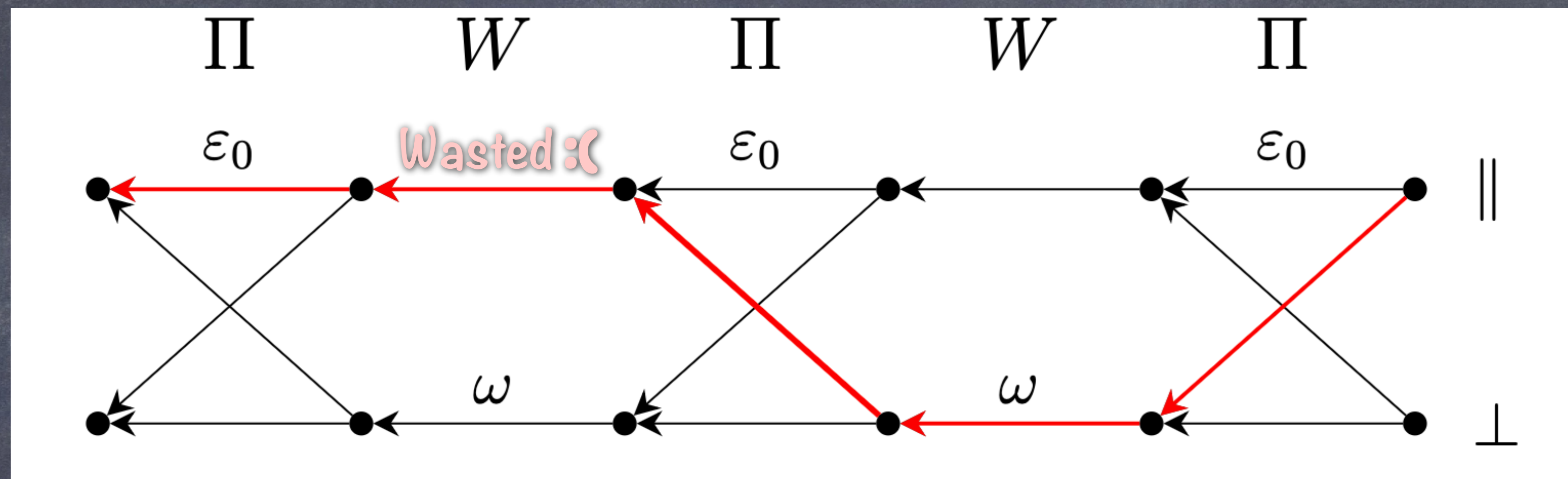
Back to reality

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Informally, every other step may be wasted, so the bound is roughly

$$\varepsilon_0^t + \omega^{t/2}$$

This 2-loss $\times$ the Ramanujan 2-“loss” yields set size $n \cdot \tilde{O}(\varepsilon^{-4})$.

TaShma's Idea

Protect your walk from the adversary

Use a fresh register at each of the t steps in a way that guarantees bias

$$\varepsilon_0^t + \omega^{t-O(1)}$$

Based on [BenAroya-TaShma STOC 2008]'s wide-replacement product.

Since the set size increases **exponentially** in t , optimizing the tradeoff yields TaShma's bound.

Our Contribution

Reuse registers

By reusing registers in a carefully chosen order, we can take length- t walks using only $O(\log t)$ registers, while still achieving $\varepsilon_0^t + \omega^{t-O(1)}$.

Since the register overhead now increases only **polynomially** in t , optimizing the tradeoff gives our improved bound.

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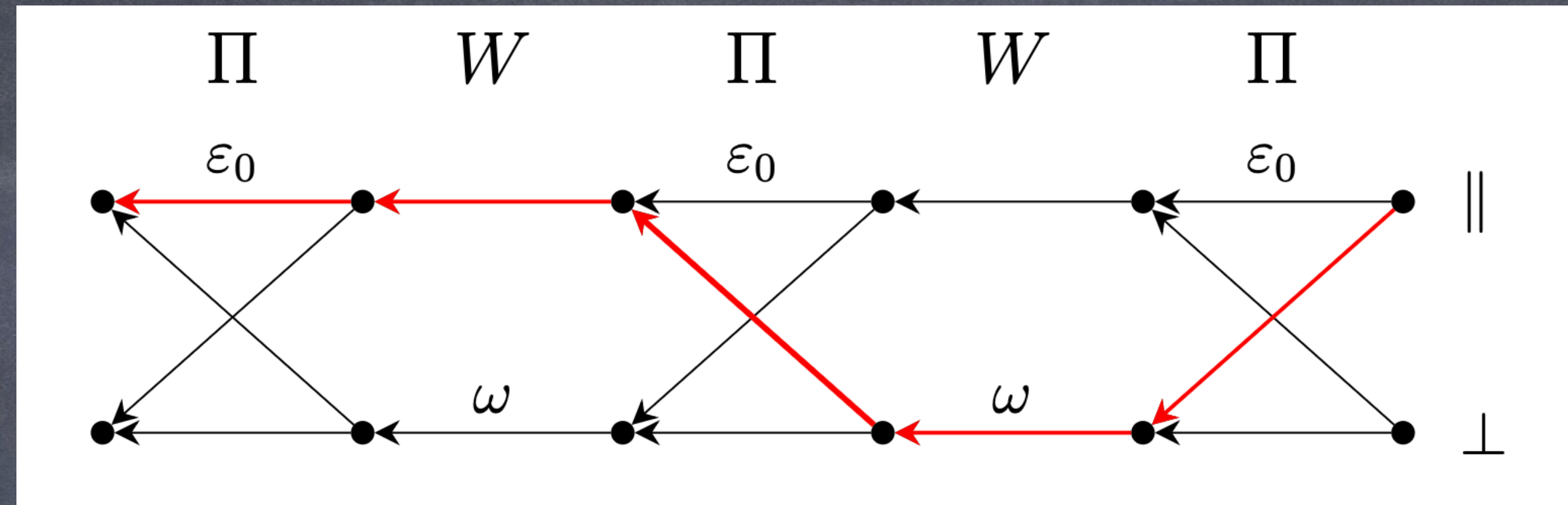
We use a Gray-code-style reuse schedule

1 2 3 1

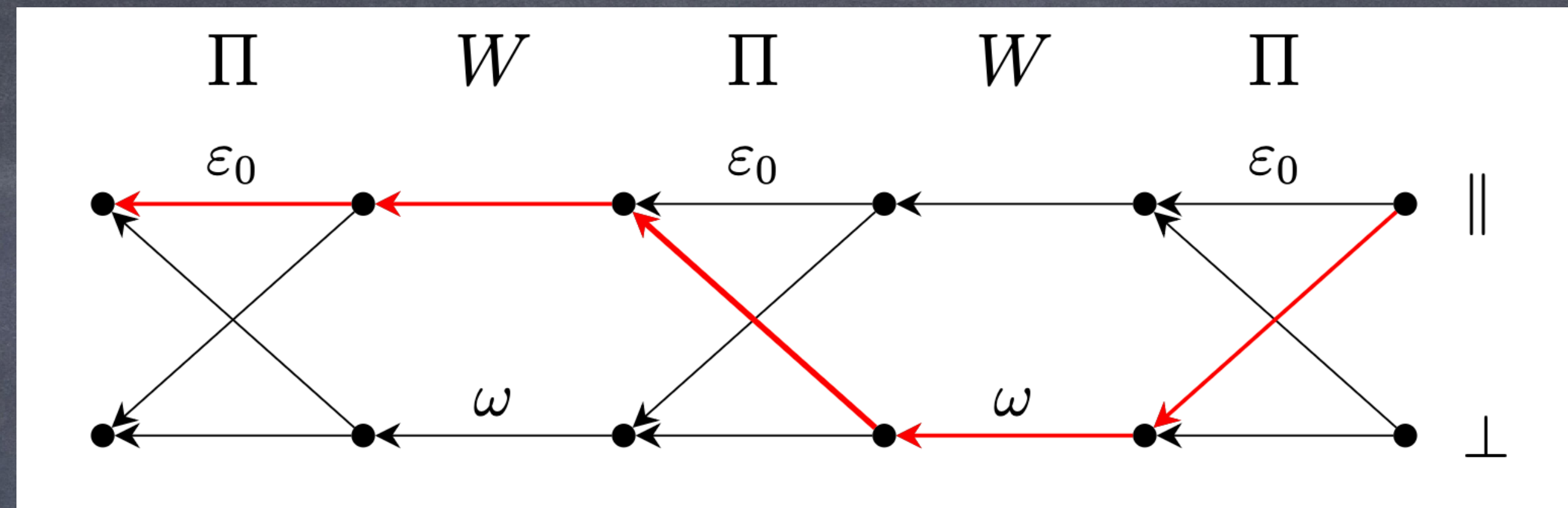
1 2 3 1 4 5 1 2 3 1

1 2 3 1 4 5 1 2 3 1 6 7 1 2 3 1 4 5 1 2 3 1

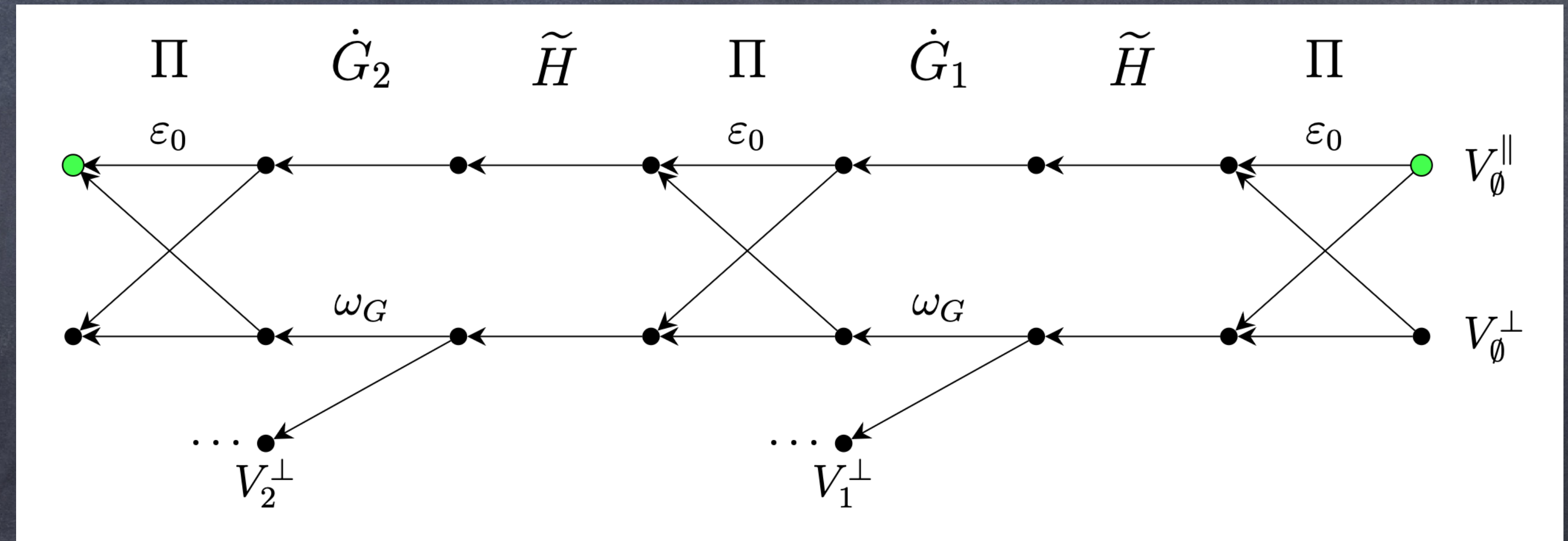
Standard walks

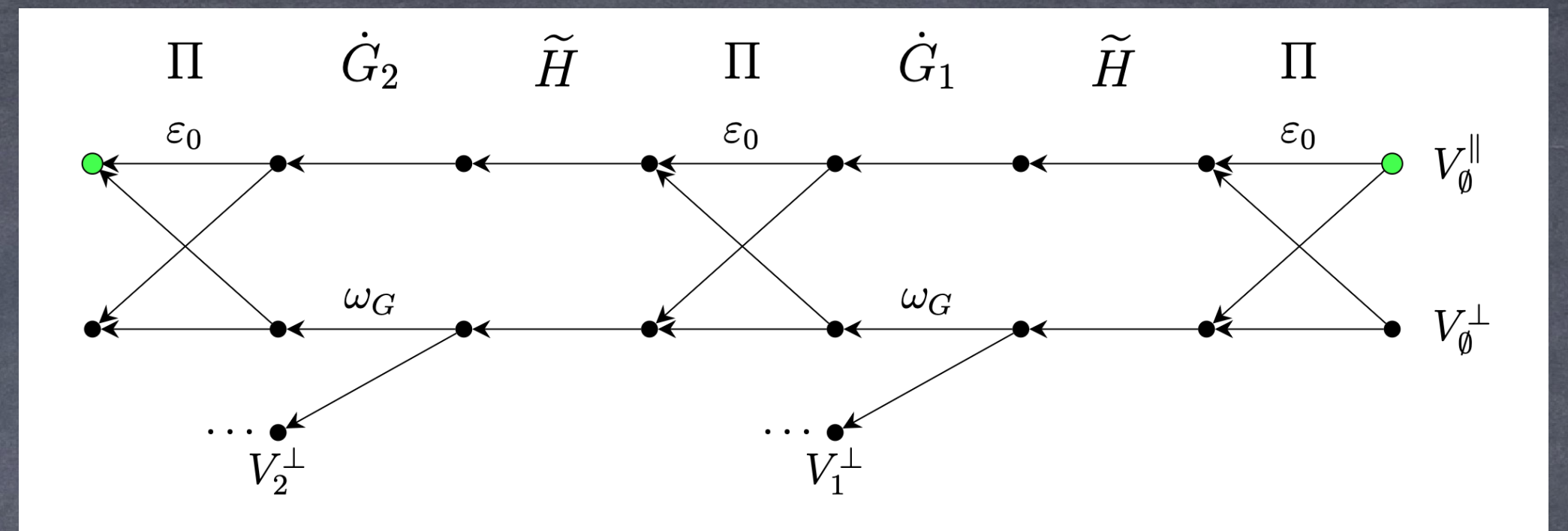
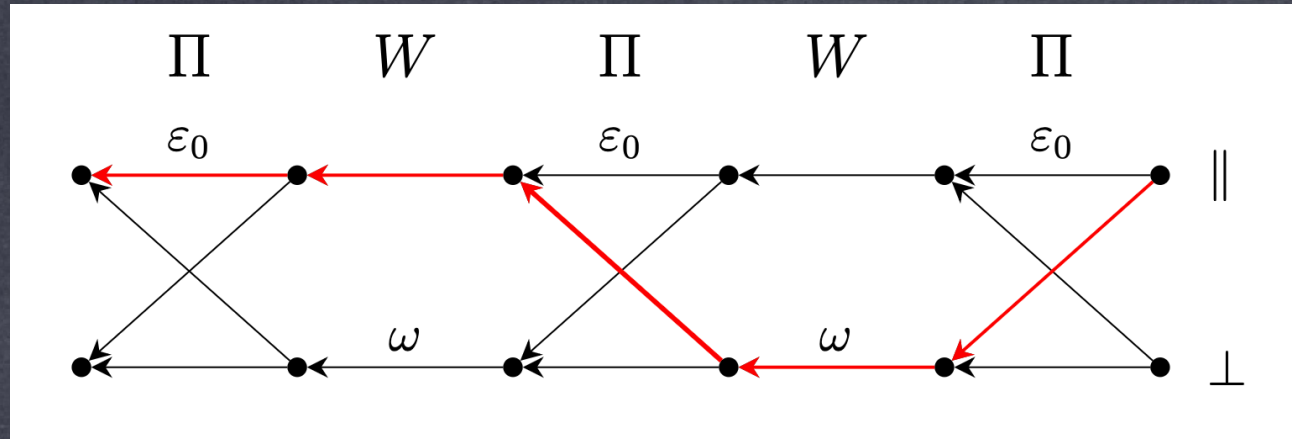


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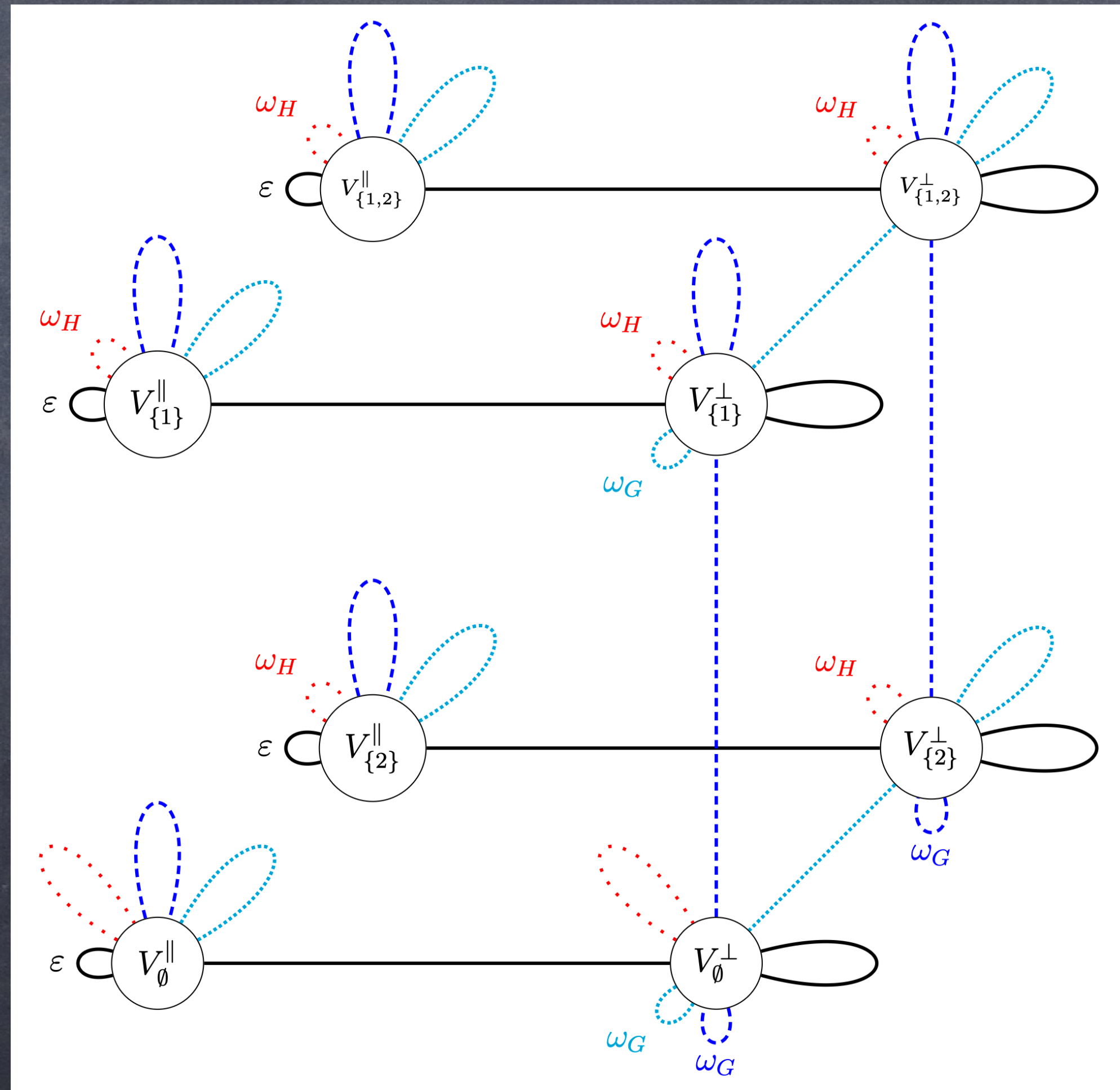


Wide walks;
fresh registers





Wide walks;
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The Second Barrier

The remaining barrier

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Ways around this barrier

- * Directed graphs can have spectral radius $\omega \approx 1/\sqrt{d}$
- * Non-backtracking walks on undirected graphs have a **linear** blowup in t .
- * Same for random walks on rotating expanders [C-Maor STOC 2023].

The challenge is to combine these with our framework.

Open problems

- * Construct explicit ε -biased sets $S \subset \{0,1\}^n$ of size $n \cdot \tilde{O}(\varepsilon^{-2})$.
- * Which bound is tight—the known lower bound or the probabilistic upper bound?
- * Can our register-reuse technique be applied elsewhere?

Thanks!