

Finding Plans and Heuristics with Spectral Graph Theory

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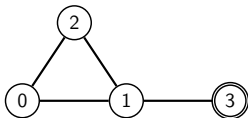


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Undirected Simple Graphs + Goal $G = \langle V, E, g \rangle$



Matrices for the Graph

$$\text{Adjacency } A = \begin{matrix} & \begin{matrix} 0 & 1 & 2 & 3 \end{matrix} \\ \begin{matrix} 0 \\ 1 \\ 2 \\ 3 \end{matrix} & \begin{bmatrix} & 1 & 1 & \\ 1 & & 1 & 1 \\ 1 & 1 & & \\ 1 & & & \end{bmatrix} \end{matrix}$$

$$\text{Degree } D = \begin{matrix} & \begin{matrix} 0 & 1 & 2 & 3 \end{matrix} \\ \begin{matrix} 0 \\ 1 \\ 2 \\ 3 \end{matrix} & \begin{bmatrix} 2 & & & \\ & 3 & & \\ & & 2 & \\ & & & 1 \end{bmatrix} \end{matrix}$$

$$\text{Laplacian } L = D - A = \begin{matrix} & \begin{matrix} 0 & 1 & 2 & 3 \end{matrix} \\ \begin{matrix} 0 \\ 1 \\ 2 \\ 3 \end{matrix} & \begin{bmatrix} 2 & -1 & -1 & \\ -1 & 3 & -1 & -1 \\ -1 & -1 & 2 & \\ -1 & & & 1 \end{bmatrix} \end{matrix}$$

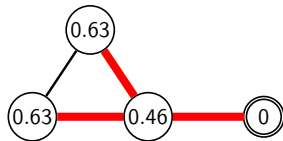
$$\text{Dirichlet } L_g = \begin{matrix} & \begin{matrix} 0 & 1 & 2 & 3 \end{matrix} \\ \begin{matrix} 0 \\ 1 \\ 2 \\ 3 \end{matrix} & \begin{bmatrix} 2 & -1 & -1 & \\ -1 & 3 & -1 & -1 \\ -1 & -1 & 2 & \\ -1 & 1 & 1 & 1 \end{bmatrix} \end{matrix}$$

Spectral Algorithm for Finding Plans [Steinerberger, 2021]

1. Compute eigenvector $\mathbf{v} \in \mathbb{R}^{|V|-1}$ such that

$$L_g \mathbf{v} = \lambda_0 \mathbf{v}$$

2. Use v_i as heuristic for vertex i
3. Follow greedily



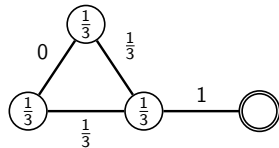
This will reach the goal!

their proof “*follows classical arguments from the continuous setting*”

It is a Solution to the Occupation Measure LP

Looks like Operator Counting Net-change Heuristic LP

$$\begin{aligned} \min_{\mathbf{x}} \quad & \sum_{\langle i,j \rangle \in E} x_{i,j} \quad \text{s.t.} \\ & out(i) - in(i) = \alpha(i) \quad \forall i \in V \setminus \{g\} \\ & in(g) = 1 \\ & x_{i,j} \geq 0 \quad \forall \langle i,j \rangle \in E \end{aligned}$$



Feasible solution induces proper policy (reaches goal with probability 1)
[Puterman, 2005]

For us 🖐️ feasible solution induces plans

Theorem

\mathbf{v} is a feasible solution for the OM LP.

Proof.

Sketch. $L_g \mathbf{v} \in \mathbb{R}^{V \setminus \{g\}}$ such that

$$(L_g \mathbf{v})_i = \underbrace{\sum_{\substack{\langle i,j \rangle \in E \\ i \prec j}} \underbrace{v_i - v_j}_{x_{i,j}}}_{out(i)} - \underbrace{\sum_{\substack{\langle i,j \rangle \in E \\ j \prec i}} \underbrace{v_j - v_i}_{x_{j,i}}}_{in(i)} = \underbrace{\lambda_0 v_i}_{\alpha(i)}$$



Corollary

By following \mathbf{v} greedily you follow a proper policy and are guaranteed to reach the goal.

Let $h(i) = \kappa v_i$ s.t. $\kappa(v_i - v_j) \leq 1 \quad \forall \langle i, j \rangle \in E$ Note $\kappa \geq 1$

Theorem

h satisfies

$$\begin{aligned} h(i) &\leq 1 + h(j) \quad \forall \langle i, j \rangle \in E \\ h(g) &= 0 \end{aligned}$$

Proof.

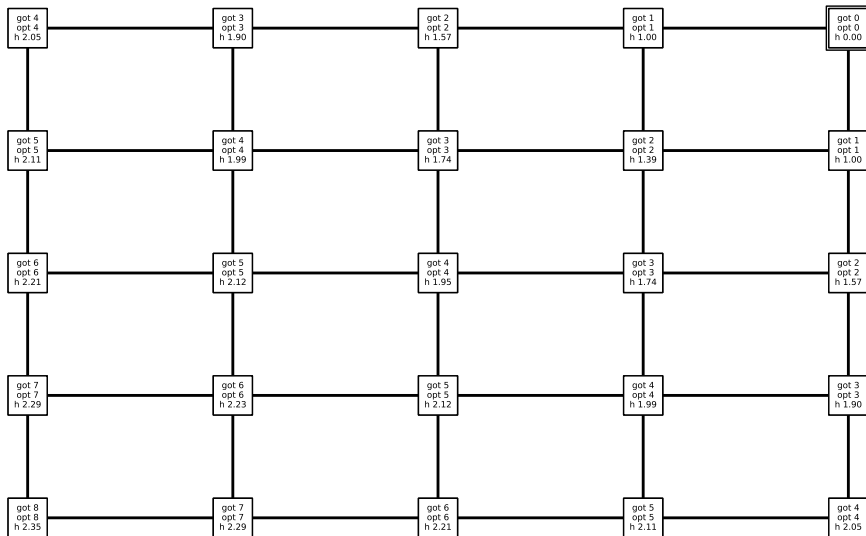
Sketch.

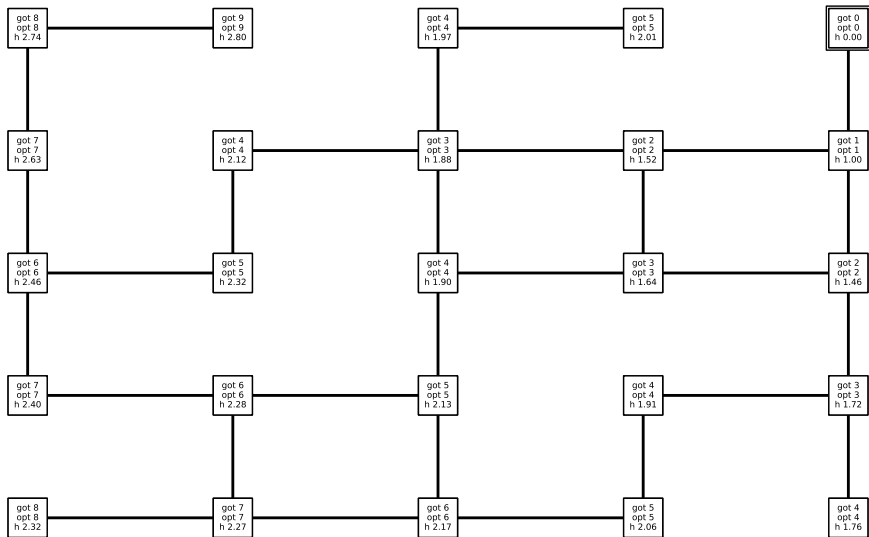
$$h(i) = \kappa v_i = \kappa(v_i - v_j) + \kappa v_j \leq 1 + h(j)$$

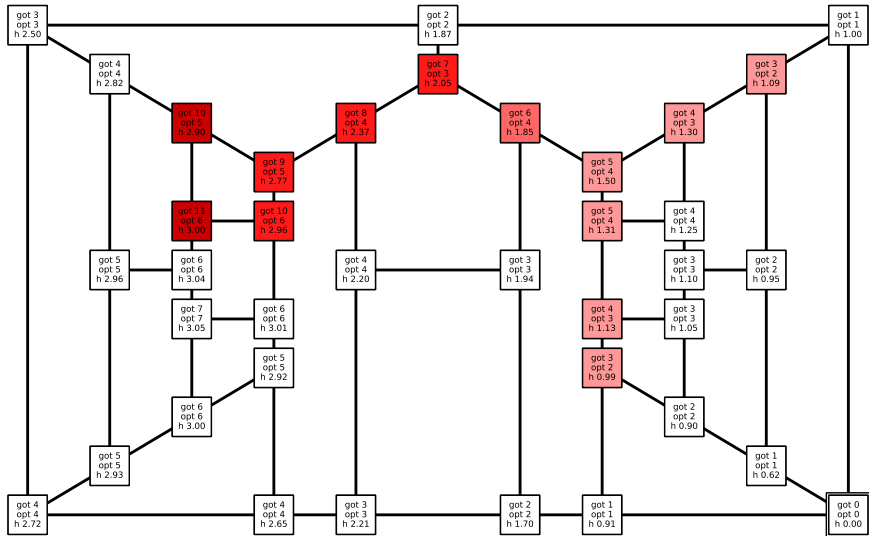


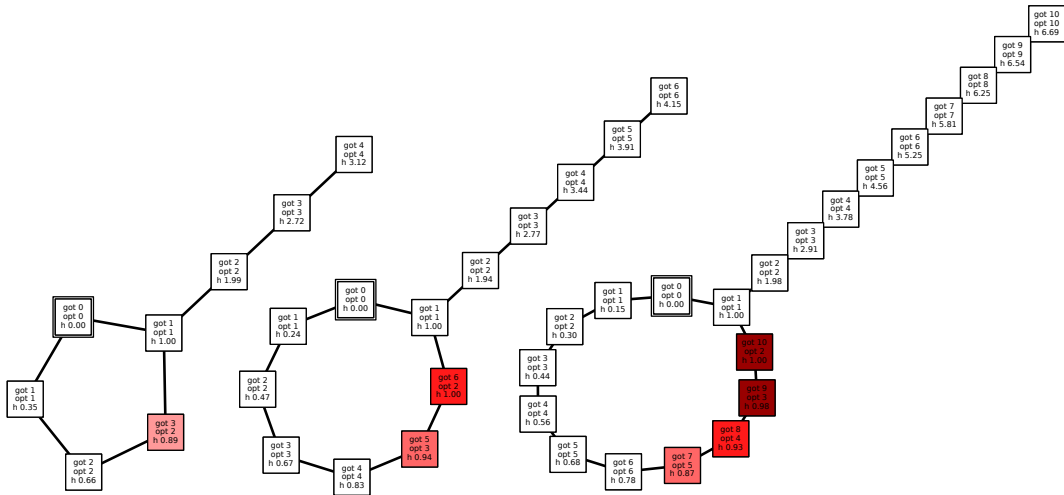
Corollary

h is a consistent, goal-aware heuristic. [Pommerening et al., 2015]



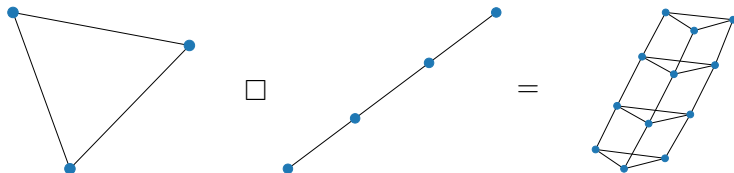






Getting transition system's matrix and finding an eigenvector is not realistic. . .

Graph Operations that Preserve Eigenvectors



if G has eigenvectors $\mathbf{x}_1, \dots, \mathbf{x}_m$ and H has eigenvectors $\mathbf{y}_1, \dots, \mathbf{y}_m$ then

$G \square H$ has eigenvectors $\mathbf{x}_i \otimes \mathbf{y}_j$ for each i, j

But \square does not give us much power. . .

Summary

- ① We can find plans with the spectral method
- ② This works because it finds a feasible solution to the Occupation Measure LP
- ③ The eigenvector also induces a consistent, goal-aware heuristic
- ④ Steinerberger [2021] suggests sticking with Dijkstra for now. . .

What's Next?

Would be nice to find a way to combine graphs that approximately preserves the required eigenvector. . .

Pommerening, F.; Helmert, M.; Röger, G.; and Seipp, J. 2015. From Non-Negative to General Operator Cost Partitioning. *Proc. of AAAI Conf. on Artificial Intelligence*, 3335–3341.

Puterman, M. L. 2005. *Markov Decision Processes*. Wiley-Interscience.

Steinerberger, S. 2021. A spectral approach to the shortest path problem. *Linear Algebra and its Applications*, 620: 182–200.