

Combinatorial optimization using Hopfield Net

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Introduction and Motivation

Optimization Problem

- What is an “Optimization Problem” ?
- Many of them are NP-Complete (Hard in some sense)
- Finding an optimum solution is tough
- Finding a 'good' solution may be easy

Why Hopfield Nets

- Optimization problem maximizes/minimizes some quantity
- Hopfield Net minimizes energy
- This motivates us to model (and possibly solve) the optimization problem using Hopfield Net

How to solve the problem using Hopfield Net

- Choose the representation such that output of the neurons corresponds to the solution of the problem
- Choose the energy function for the problem such that its minimum value gives the best solution to the problem
- Compare the energy equation of the problem E_p to the energy equation of the Hopfield Net E_n
- Find the weight and threshold values

Agenda

Revision

- Travelling Salesman Problem
- Sorting

Problems in Graph Theory

- Vertex Cover
- Independent Set
- Clique
- Matching

Travelling Salesman Problem

Energy Equations

$$E_1 = \frac{A}{2} \left[\sum_i (\sum_\alpha x_{i,\alpha} - 1)^2 + \sum_\alpha (\sum_i x_{i,\alpha} - 1)^2 \right]$$

$$E_2 = \frac{1}{2} \left[\sum_i \sum_j \sum_\alpha d_{ij} x_{i\alpha} (x_{j,\alpha+1} + x_{j,\alpha-1}) \right]$$

$$E_p = E_1 + E_2$$

$$E_n = - \sum_i \sum_\alpha \sum_j \sum_\beta w_{i\alpha,j\beta} x_{i\alpha} x_{j\beta} - \sum_i \sum_\alpha x_{i\alpha} \theta_{i\alpha}$$

Weight and threshold values

Row Weight $w_{ij} = -A$

Column Weight $w_{ij} = -A$

Cross Weight $w_{ij} = - \left(\frac{d_{ij} + d_{ji}}{2} \right)$

$\theta_{ij} = 2A$

Sorting

Representation

Create a neuron corresponding to i^{th} value and j^{th} position
 If the neuron fires, i^{th} value will be in j^{th} position

Equations

- A value can be in at most 1 position
- A position can accommodate at most 1 value
- Summation over *value* * *position* should be minimized

$$E_1 = \frac{A}{2} \left[\sum_i (\sum_\alpha x_{i,\alpha} - 1)^2 + \sum_\alpha (\sum_i x_{i,\alpha} - 1)^2 \right]$$

$$E_2 = B \sum_i \sum_\alpha x_{i\alpha} * S_i * \alpha$$

E_p and E_n

$$E_p = \frac{A}{2} \left[\sum_i (\sum_\alpha x_{i,\alpha} - 1)^2 + \sum_\alpha (\sum_i x_{i,\alpha} - 1)^2 \right] + B \sum_i \sum_\alpha x_{i\alpha} * S_i * \alpha$$

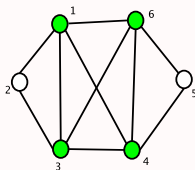
$$E_n = - \sum_i \sum_\alpha \sum_j \sum_\beta w_{i\alpha,j\beta} x_{i\alpha} x_{j\beta} - \sum_i \sum_\alpha x_{i\alpha} \theta_{i\alpha}$$

Weight and threshold values

$$w_{ij} = -A$$

$$\theta = 2A - S_i B \alpha$$

Vertex Cover (VC)



Representation

Create a neuron for each vertex

If neuron fires, the vertex will be in the VC

Equations

- Minimize the no. of vertices in the VC

$$E_1 = \sum_i v_i$$

- Cover each edge

$$E_2 = \sum_i \sum_j d_{ij} \overline{v_i \vee v_j}$$

Simplified energy equation

$$\begin{aligned}
 E_p &= A \sum_i \sum_j d_{ij} \overline{v_i \vee v_j} + B \sum_i v_i \\
 &= A \sum_i \sum_j d_{ij} (1 - v_i \vee v_j) + B \sum_i v_i \\
 &= A \sum_i \sum_j d_{ij} (1 - v_i - v_j + v_i v_j) + B \sum_i v_i \\
 &= A \sum_i \sum_j d_{ij} v_i v_j - A \sum_i \sum_j d_{ij} (v_i + v_j) + B \sum_i v_i + A \sum_i \sum_j d_{ij}
 \end{aligned}$$

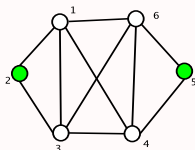
 E_p and E_n

$$\begin{aligned}
 E_p &= A \sum_i \sum_j d_{ij} v_i v_j - A \sum_i \sum_j d_{ij} (v_i + v_j) + B \sum_i v_i \\
 E_n &= - \sum_i \sum_j w_{ij} x_i x_j - \sum_i x_i \theta_i
 \end{aligned}$$

Weight and threshold values

$$\begin{aligned}
 w_{ij} &= -2A d_{ij} \\
 \theta_{ij} &= 2A \sum_i d_{ij} - B
 \end{aligned}$$

Independent set (IS)



Representation

Create a neuron for each vertex

If neuron fires, the vertex will be in the IS

Equations

- Maximize the no. of vertices in the IS

$$E_1 = \sum_i \bar{v}_i$$

- There shouldn't be an edge between the vertices in the IS

$$E_2 = \left[\sum_i \sum_j e_{ij} (v_i \wedge v_j) \right]$$

Simplified energy equation

$$\begin{aligned}
 E_p &= A \left[\sum_i \sum_j e_{ij} (v_i \wedge v_j) \right] + B \sum_i \bar{v}_i \\
 &= A \left[\sum_i \sum_j e_{ij} (v_i v_j) \right] - B \sum_i v_i + B
 \end{aligned}$$

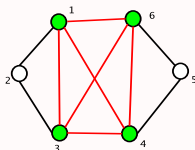
 E_p and E_n

$$\begin{aligned}
 E_p &= A \left[\sum_i \sum_j e_{ij} (v_i v_j) \right] - B \sum_i v_i \\
 E_n &= - \sum_i \sum_j w_{ij} x_i x_j - \sum_i x_i \theta_i
 \end{aligned}$$

Weight and threshold values

$$\begin{aligned}
 w_{ij} &= -Ae_{ij} \\
 \theta_i &= B
 \end{aligned}$$

Clique



Representation

Create a neuron for each vertex

If neuron fires, the vertex will be in the Clique

Equations

- Maximize the no. of vertices in the Clique

$$E_1 = \sum_i \bar{v}_i$$

- Confirm that there is an edge between every pair of vertices

$$E_2 = - \left[\sum_i \sum_j (e_{ij} - 1)(v_i \wedge v_j) \right]$$

Simplified energy equation

$$\begin{aligned}
 E_p &= -A \left[\sum_i \sum_j (e_{ij} - 1)(v_i \wedge v_j) \right] + B \sum_i \bar{v}_i \\
 &= -A \left[\sum_i \sum_j (e_{ij} - 1)(v_i v_j) \right] - B \sum_i v_i + B
 \end{aligned}$$

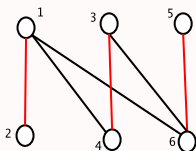
 E_p and E_n

$$\begin{aligned}
 E_p &= -A \left[\sum_i \sum_j (e_{ij} - 1)(v_i v_j) \right] - B \sum_i v_i \\
 E_n &= - \sum_i \sum_j w_{ij} x_i x_j - \sum_i x_i \theta_i
 \end{aligned}$$

Weight and threshold values

$$\begin{aligned}
 w_{ij} &= Ae_{ij} - 1 \\
 \theta_i &= B
 \end{aligned}$$

Matching



Representation

Create a neuron for each edge

If neuron fires, the edge will be in the Matching

Equations

- Maximize the no. of edges in the Matching

$$E_1 = \sum_i \bar{e}_i$$

- The edges in the Matching shouldn't be adjacent

$$E_2 = \left[\sum_i \sum_j d_{ij} (e_i \wedge e_j) \right]$$

Simplified energy equation

$$\begin{aligned}
 E_p &= A \left[\sum_i \sum_j d_{ij} (e_i \wedge e_j) \right] + B \sum_i \bar{e}_i \\
 &= A \left[\sum_i \sum_j d_{ij} (e_i e_j) \right] - B \sum_i e_i + B
 \end{aligned}$$

 E_p and E_n

$$\begin{aligned}
 E_p &= A \left[\sum_i \sum_j d_{ij} (e_i e_j) \right] - B \sum_i e_i \\
 E_n &= - \sum_i \sum_j w_{ij} x_i x_j - \sum_i x_i \theta_i
 \end{aligned}$$

Weight and threshold values

$$\begin{aligned}
 w_{ij} &= -A d_{ij} \\
 \theta_i &= B
 \end{aligned}$$

Conclusion

- mathematical formulation for the problems (requires ingenuity)
- intuitive values for weights and thresholds
- cute programs

References

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