

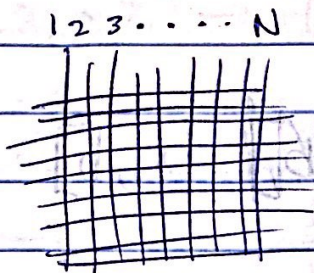
HBC:

uses noise to escape lesser minima

Simulated Annealing

→ traveling salesperson
(info traveling through a network/path, what's the fastest/best route)

N cities



A_{ij} is the cost of traveling from city i to city j

Properties of Symmetric Matrices

orthogonal vectors

1) $u^T v = u \cdot v = 0$ (dot prod is 0)

2) $\{q_1, q_2, \dots, q_n\}$ $q_i^T q_j = 0$, $q_i \in \mathbb{R}^n$
 $\vec{x} = \sum_{j=1}^n c_j q_j$ for any $\vec{x} \in \mathbb{R}^n$

diagonals must be real

ex $A = \begin{bmatrix} 1 & 2i \\ 2-i & 1 \end{bmatrix}$

$A \in \mathbb{R}^{N \times N}$ and symmetric

OR $A \in \mathbb{C}^{N \times N}$ and Hermitian

$A = A^* = (\overline{A^T})$
(transpose conjugate)

1) A has real eigenvalues

2) all of A 's eigenvectors are orthogonal

~~$(Ax) \cdot (y) = (x \cdot (A^*y))$~~

Assume $A \in \mathbb{C}^{N \times N}$ and Hermetian ($A = A^*$)

Prove (1) $AV = \lambda V$
 $V^T A V = \lambda V^T V$ left multiply by V^T
 $V^* A V = \lambda V^* V$ (or V^* b/c V could have complex #s)
 $\lambda = \frac{V^* A V}{V^* V}$ (& transpose)

* try to conjugate both sides \rightarrow if it's the same, λ is real! *

$$\lambda^* = \left(\frac{V^* A V}{V^* V} \right)^*$$

* same Transpose rules apply to * \star $(BC)^T = C^T B^T$

$$\lambda^* = \frac{V^* A^* V^{**}}{V^* V^{**}} = \frac{V^* A^* V}{V^* V}$$

b/c $A^* = A$ (A is Hermetian),

$$\lambda^* = \frac{V^* A V}{V^* V}$$

So, λ is real!

Prove (2)

$$AV = \lambda V$$

$$AX = \xi X$$

Assume ① $\lambda \neq \xi$, ② $V^* X \neq 0$
 so $V \neq X$ b/c e-vec are LI

$$A(V-X) = \lambda V - \xi X$$

$$(V-X)^* A(V-X) = (V-X)^* (\lambda V - \xi X)$$

$$\underbrace{V^* A V}_{\lambda} - V^* A X - X^* A V + \underbrace{X^* A X}_{\xi} = \lambda V^* V - \lambda X^* V - \xi V^* X + \xi X^* X$$

$$V^* A X - X^* A V = -\lambda X^* V - \xi V^* X$$

change A to A^*

$$-V^* A^* X - X^* A^* V = -\lambda X^* V - \xi V^* X$$

$$+ \lambda V^* X + \xi X^* V = +\lambda X^* V + \xi V^* X$$

Since $\lambda \neq \xi$, $V^* X = 0$. Contradiction

Since $(AV)^* = (A^* V)^*$
 $V^* A^* = \lambda V^*$
 (same for 2nd eqn)

Krylov Subspace

"find the best soln' of problem in a reduced subspace & keep building the space up"

- Lanczos Algorithm: eigenvalue

- GMRES: $Ax=b$

- Conjugate Gradient

$$\frac{v^* A^* v}{v^* v} = \frac{v^* A v}{v^* v} = \lambda$$

$$A = A^*$$

$$v A^* v = v A v$$

$$v A v = \lambda v^* v$$

$$A v = \lambda v$$

$$A^* v = \lambda v$$

$$(x^S - v^S)^* (x - v) = (x - v)^* A^* (x - v)$$

$$(x^S - v^S)^* (x - v) = (x - v)^* A^* (x - v)$$

$$x^S + x^* v^S - v^S x^* - (v^S)^* v^S = x A^* x + v A^* x - x A^* v - v A^* v$$

$$x^S = \frac{1}{x^* x} x^* x^S$$

$$x^* v^S - v^S x^* = v A^* v - v A^* v$$

$$x^* v^S - v^S x^* = v A^* v - v A^* v$$

$$x^* v^S - v^S x^* = v A^* x - x^* A^* v$$