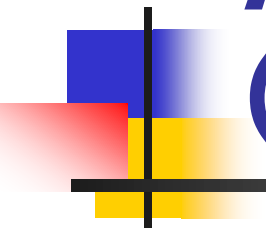

Artificial Intelligence Methods (G52AIM)

Dr Rong Qu

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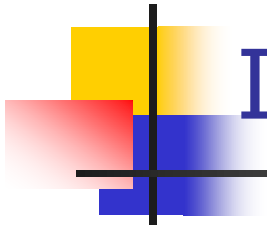
Variable Neighborhood Search



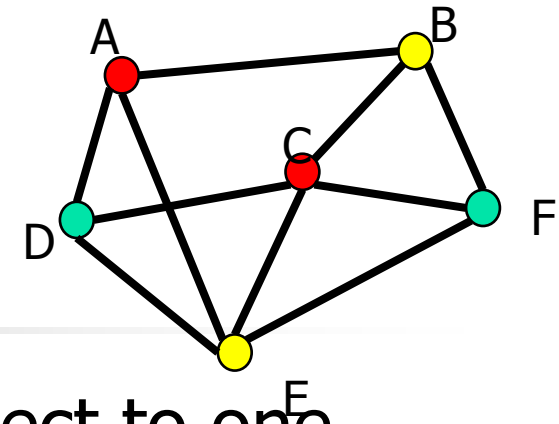
Idea of VNS

- So far on our local search based approaches, only one neighbourhood is employed

- To escape from local optimum
  - In SA, move to worse neighbourhoods base on a probability using the cooling schedule
  - In TS, move to worse neighbourhoods which are not tabued

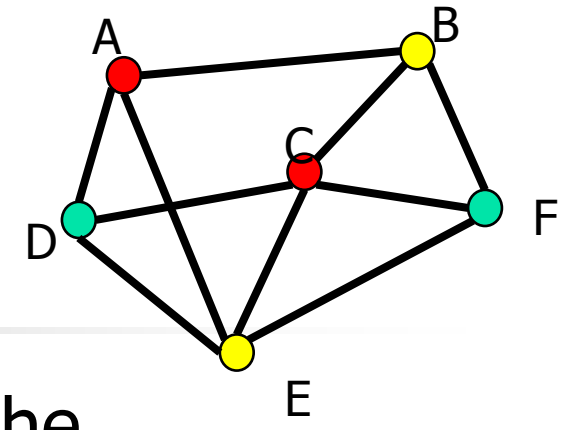
  - Any other way of escape from local optimum?

Idea of VNS



- **Fact 1.** A local minimum with respect to one neighbourhood structure is not necessary so for another;
- **Fact 2.** A global minimum is a local minimum with respect to all possible neighbourhood structures;
- **Fact 3.** For many problems local minima with respect to one or several neighbourhoods are relatively close to each other.

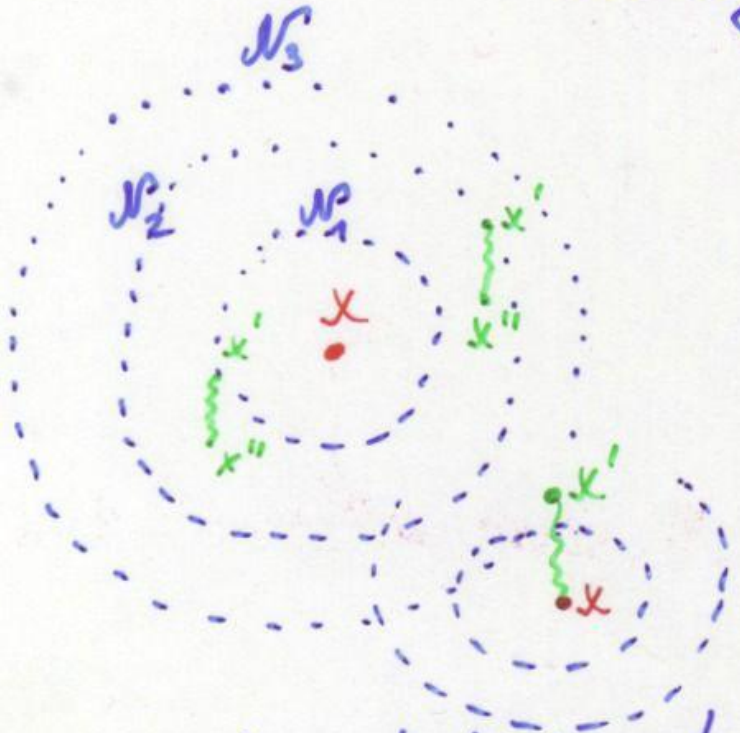
Idea of VNS



- **Principle**: systematically change the neighbourhood during the search

- Notation
  - N_k, $k = 1, 2, \ldots k_{max}$, is the set of neighbourhood structures
  - $N_k(s)$ is the set of solutions in the k^{th} neighbourhood of incumbent solution s

Basic idea: • Escape from local optima trap by changing the neighborhood structure.



• Only global opt. is local for any $\mathcal{N}_k$

Basic VNS

• Select a set of neighborhoods $\mathcal{N} = \{\mathcal{N}_1, \ldots, \mathcal{N}_{k_{max}}\}$

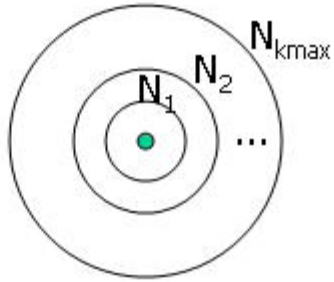
• Do while stop
  k = 1
  Do while ($k < k_{max}$)
    • Shaking
    • Localsearch
    • Move or not

Variable Neighborhood Search

Initialisation

Select the set of neighbourhood structures N_k

Find an initial solution x

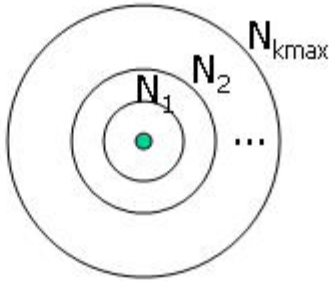


Variable Neighborhood Search

Initialisation

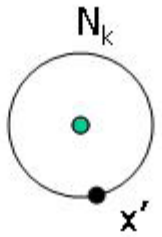
Select the set of neighbourhood structures N_k

Find an initial solution x



Repeat until stopping condition is met

- Set K=1

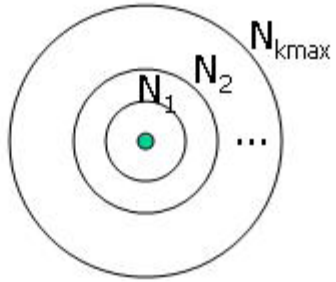


Variable Neighborhood Search

Initialisation

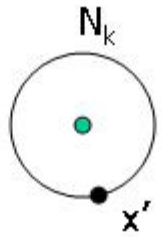
Select the set of neighbourhood structures N_k

Find an initial solution x



Repeat until stopping condition is met

- Set K=1
- **Repeat** until $k=k_{max}$
  1. *Shaking*: Generate a random point X′ in $N_k(x)$

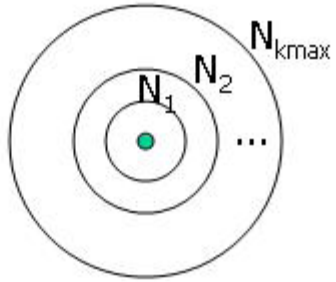


Variable Neighborhood Search

Initialisation

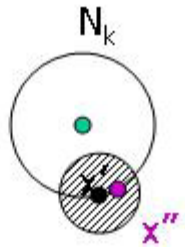
Select the set of neighbourhood structures N_k

Find an initial solution x



Repeat until stopping condition is met

- Set K=1
- **Repeat** until $k=k_{max}$
  1. *Shaking*: Generate a random point X′ in $N_k(x)$
  2. *Local Search*: x″ is the obtained optimum

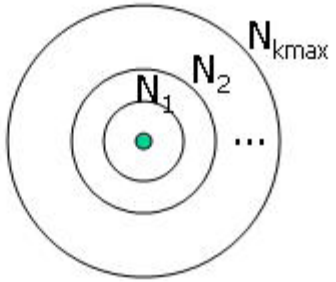


Variable Neighborhood Search

Initialisation

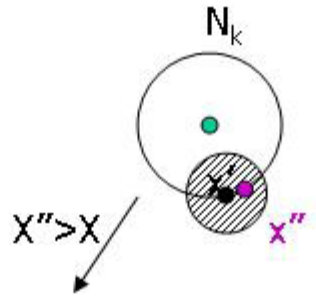
Select the set of neighbourhood structures N_k

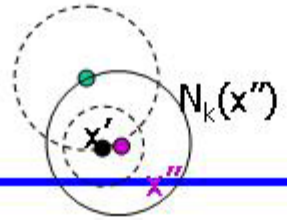
Find an initial solution x



Repeat until stopping condition is met

- Set K=1
- **Repeat** until $k=k_{max}$
  1. *Shaking*: Generate a random point X′ in $N_k(x)$
  2. *Local Search*: x″ is the obtained optimum
  3. Move or not:
     - If x″ is better than x then x=x″ and k=1



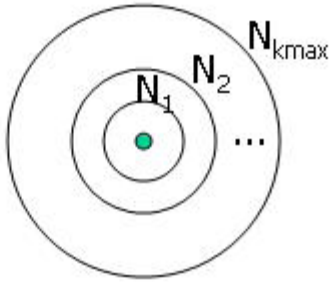


Variable Neighborhood Search

Initialisation

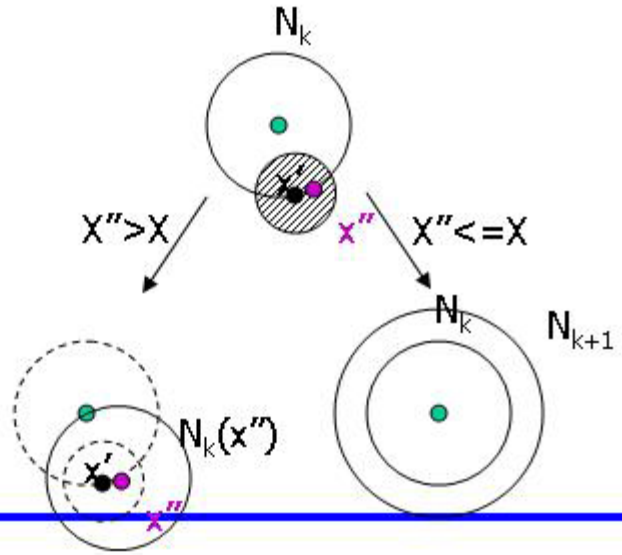
Select the set of neighbourhood structures N_k

Find an initial solution x



Repeat until stopping condition is met

- Set K=1
- **Repeat** until $k=k_{max}$
  1. *Shaking*: Generate a random point X′ in $N_k(x)$
  2. *Local Search*: x″ is the obtained optimum
  3. Move or not:
     - If x″ is better than x then x=x″ and k=1
     - Otherwise k=k+1

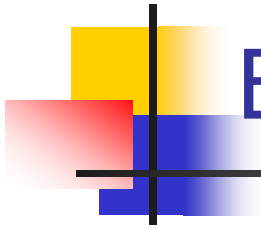


Basic VNS

Procedure Basic VNS

Select the neighbourhood set; find an initial solution,
choose a stopping condition
Repeat until stopping condition is met:
set the first neighborhood
Repeat until the last neighbourhood
Shaking
Local Search
Move or Not

End



Basic VNS

- Shaking
  - Generate a point x' at random from the kth neighbourhood of x
- Local Search
  - Apply some local search method with x' as initial solution
  - x'' be the obtained local search
- Move or Not
  - If x'' is better than x, move to x'', go to the first neighbourhood
  - Otherwise go to the next neighbourhood

There was a typo in the original slide, i.e. x' instead of x. It should be x here, storing the incumbent solution during the search.

Basic VNS

Procedure Basic VNS

Select $\{N_k\}$, $k = 1, \ldots, k_{max}$; find an initial solution x, choose a stopping condition

Repeat until stopping condition is met:

$k \leftarrow 1$

Repeat until $k = k_{max}$

$x' \leftarrow$ RandomSolution$(N_k(x))$

$x'' \leftarrow$ LocalSearch(x')

if $f(x'') < f(x)$ then

$x \leftarrow x''$

$k \leftarrow 1$

else

$k \leftarrow k + 1$

until stopping condition

End

Basic VNS – variants

- Order of the neighbourhoods
  - forward VNS: start with $k = 1$ and increase k by one if no better solution is found; otherwise set $k \leftarrow 1$
  - backward VNS: start with $k = k_{max}$ and decrease k by one if no better solution is found
  - extended version: parameters k_{min} and k_{step}; set $k \leftarrow k_{min}$ and increase k by k_{step} if no better solution is found

Variable Neighbourhood Descent

Procedure VND

Select $\{N_k\}$, $k = 1, \ldots, k_{max}$, find an initial solution x

$k \leftarrow 1$

Repeat until $k > k_{max}$

$x' \leftarrow$ FindBestNeighbour(x)

if $f(x') < f(x)$ then

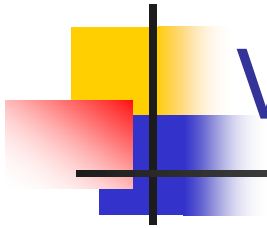
$x \leftarrow x'$

$k \leftarrow 1$

else

$k \leftarrow k + 1$

End



Variable Neighbourhood Descent

- Change the neighbourhood in a deterministic way
- Final solution is locally optimal w.r.t. all neighbourhoods
- First improvement may be applied instead of best improvement
- Typically, order neighbourhoods from smallest to largest

Reduced VNS

Procedure Reduced VNS

Select $\{N_k\}$, $k = 1, \ldots, k_{max}$, find an initial solution x, choose a stopping condition

$k \leftarrow 1$

Repeat until $k = k_{max}$

$x' \leftarrow$ RandomSolution$(N_k(x))$

if $f(x') < f(x)$ then

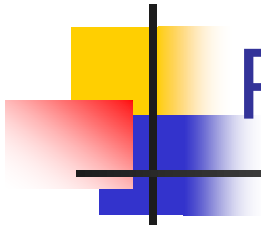
$x \leftarrow x'$

$k \leftarrow 1$

else

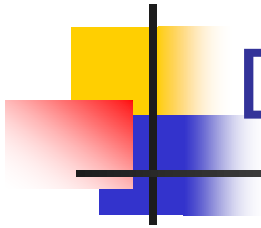
$k \leftarrow k + 1$

End



Reduced VNS

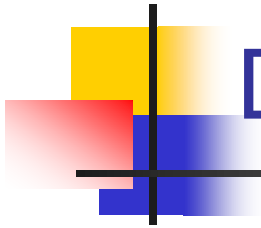
- Same as basic VNS except that no LocalSearch procedure is applied
- Only explores randomly different neighbourhoods
- Can be faster than standard local search algorithms for reaching good quality solutions



Decisions in VNS

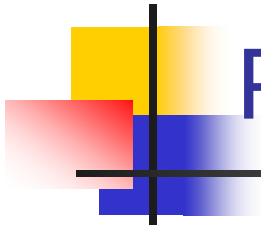
- Number and type of neighbourhoods to be used
- Order of their use in the search
- Strategy for changing the neighbourhoods
- Local search methods
- Stopping condition

- There is no need to design sophisticated acceptance criteria to escape from local optima
- Design of neighbourhoods presents to be crucial for VNS



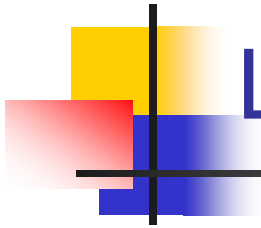
Decisions in VNS – for TSP

- Number and type of neighbourhoods to be used
- Order of their use in the search
- Strategy for changing the neighbourhoods
- Local search methods
- Stopping condition



Reading Materials

- N. Mladenovic, Variable neighbourhood search. Invited seminar, City University, London, March 8, 2007.
- P. Hansen and N. Mladenovic, Variable neighborhood search: Principles and applications, EJOR 43, 2001.



Learning Objectives

- VNS Basics
  - Basic VNS
  - Reduced VNS
  - Supporting idea
  - Decisions in algorithm design
- Be able to implement VNS in your coursework
  - Making appropriate decisions to design effective and efficient VNS for the problem in hand