

AI1: Topic 1 : Week 4

Search in continuous spaces

Attendance Code:



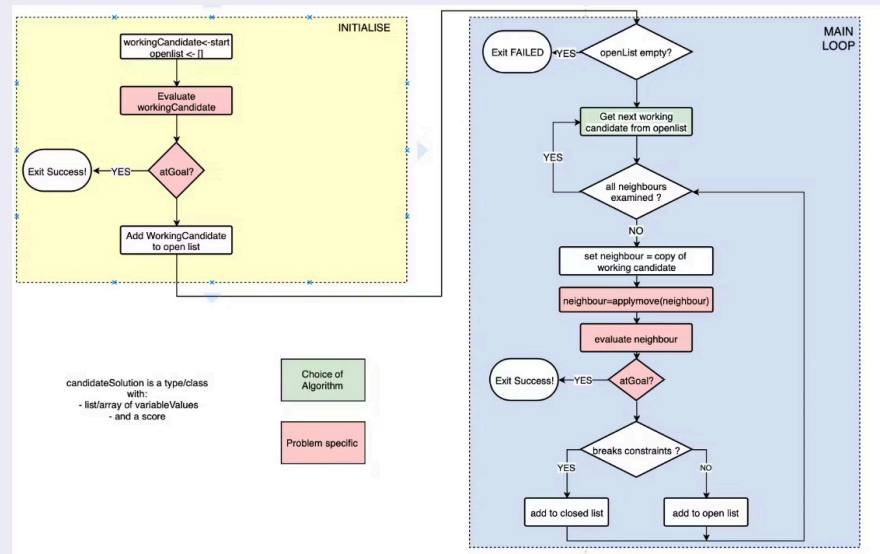
any questions about the module so far?



This week

- Recap questions and discussions
- Search in continuous spaces
- Round up of topic





flowchart for common search framework



INITIALISE

```
Set open_list, closed_list ← EmptyList
working_candidate ← Initialise (CandidateSolution)
Test ( working_candidate) Problem-specific code
AppendToOpenList(working_candidate)
```

MAIN LOOP

```
WHILE IsNotEmpty( open_list) DO
  working_candidate ← SelectAndMoveFromOpenList(algorithm_name) Algorithm-specific code
  FOR sample in SAMPLE_SIZE DO
    GENERATE
      neighbour ← ApplyMoveOperator (working_candidate) Representation-specific code
    TEST
      status ← Test ( neighbour) Problem-specific code
    UPDATE WORKING MEMORY
      IF status IS AtGoal THEN
        Return(SUCCESS)
      ELSE IF status IS BREAKS_CONSTRAINTS THEN
        AppendToClosedList(neighbour)
      ELSE
        AppendToOpenList(neighbour)
  AppendToClosedList(workingCandidate)
```

pseudocode for framework



I understand

Strongly disagree

The idea of search using a common framework

How we can 'plug in' different problems to solve them

How we apply different algorithms within the framework

Strongly agree



I understand

Strongly disagree

The flowchart of the design



The pseudocode of the design



How to work with Jim's implementation



Strongly agree



How do we select from the open list?

- Depth-first: Age (youngest first)
- Breadth-first: Age: (oldest first)
- Local Search (hill-climbing): quality first, discard rest
- Best-first: quality first, keep rest
- A*: : cost to reach + quality
- Dijkstra: Cost to reach



Which of these are *guaranteed* to find the optimal solution?

✗	✓	✗	✗	✓	✗
Depots- first	Breadth- first	Local Search	best- first	A*	Dijkstra

Hill-Climbers can get stuck in local optima



True



False

What do you understand by "optimal"?



Choosing the right algorithm for a problem

The next few questions describe some
scenarios.

In each, select which algorithms you think it
might be appropriate to apply



A 'password-cracking' problem

0	0	0	0	0
depth- first	breadth- first	local search	best-first	A Star
✓	✓	✗	✗	✗



Scheduling production in a local cider factory?






0	0	0	0	0
depth-first	breadth-first	local search	best-first	A Star
✗	✗	✓	✓	✓



A npc planning a path to chase someone in a game?

✗	✗	✗	✗	✓
depth- first	breadth- first	local search	best-first	A Star

Organising daily delivery routes ?

				
depth- first	breadth- first	local search	best-first	A Star

Leaderboard

No results yet

Top Quiz participants will be displayed here once there are results!

I'd expect you to be able to ...

- recognise what algorithm is being applied if I showed you code
- tell me what line defined the algorithm
- characterise algorithms in terms of efficiency, completeness, optimality
- characterise the **context** of a problem in terms of the trade-offs between completeness/optimality/efficiency
- select the most suitable **algorithm(s)** for a simple problem-solving scenario
- select the most suitable **representation** for a simple problem,



Searching in different types of spaces

- A candidate solution encodes values for a set of decisions that define a solution
- examples so far: decisions take **categorical** values
- so at every stage we have a fixed set of neighbours
- What happens if the decisions take real/continuous values?



Continuous variables have infinite numbers of neighbours!

only limited by precision of floating points
so we can't examine all the neighbours.

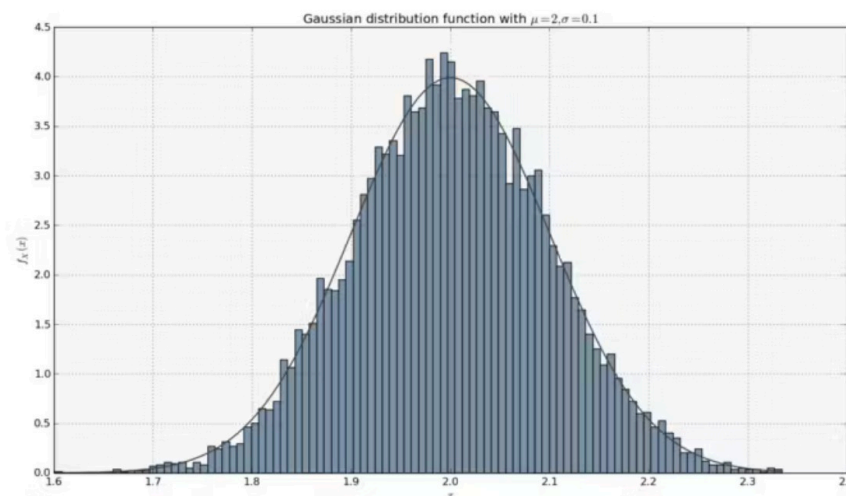
Option1: make neighbours by adding noise and
just sample a few

Option 2: put. the effort into use some cunning
maths, and just generate one neighbour



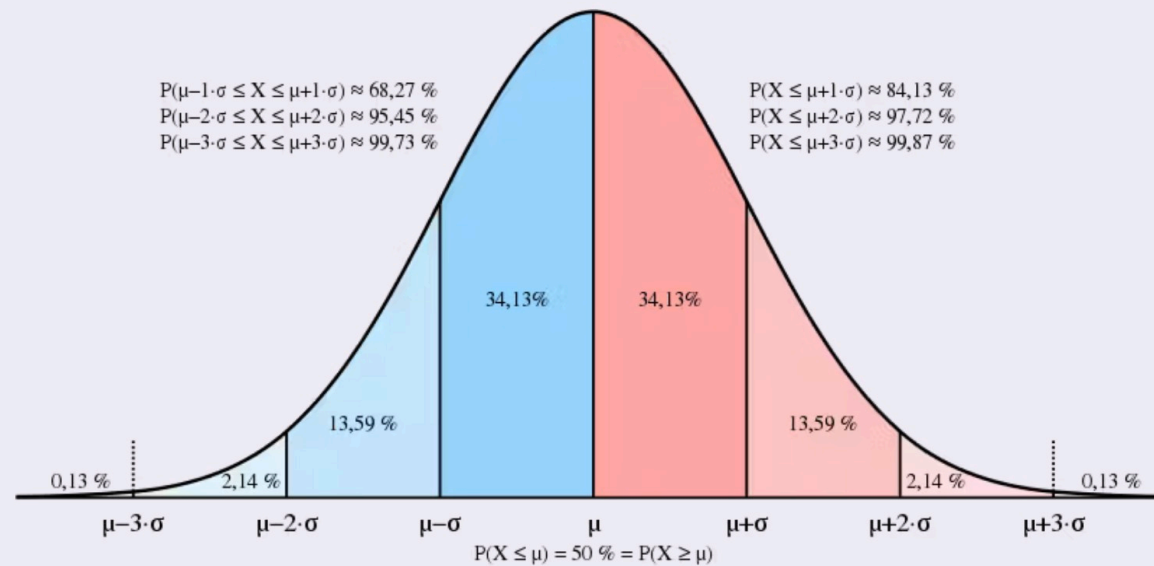
Can any one describe what a Gaussian/Normal distribution of numbers is





normal distribution from samples Mean is called mu (greek letter like a u) and standard deviation is called sigma (like an o with a tail)





Probability of generating points from a random distribution



Option 1: Adding noise and sampling

- **Zero-mean** : change up as likely as change down
- **Standard Deviation** (sigma) choose to suit scale of problem
- For each variable:
 - * generate a random number from a $N(0,1)$ distribution
 - * scale to problem (multiply by sigma)
 - * add the random numbers to the variable
- Might need some trial and error to decide how many samples to use



option 2: where possible

- Apply some maths to estimate local slope
- do this while calculating quality
- move operator = 1 step in that direction
- repeat



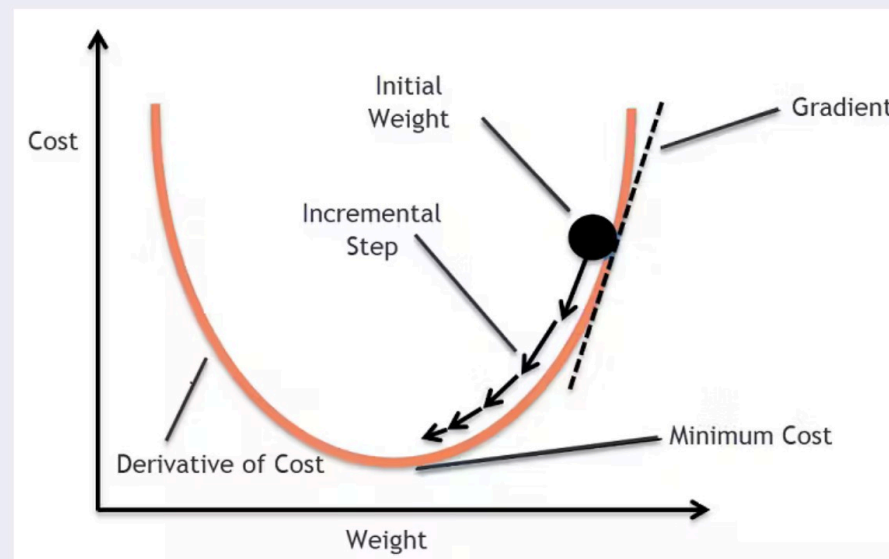


Image from <https://www.analyticsvidhya.com/blog/2020/10/how-does-the-gradient-descent-algorithm-work-in-machine-learning/>



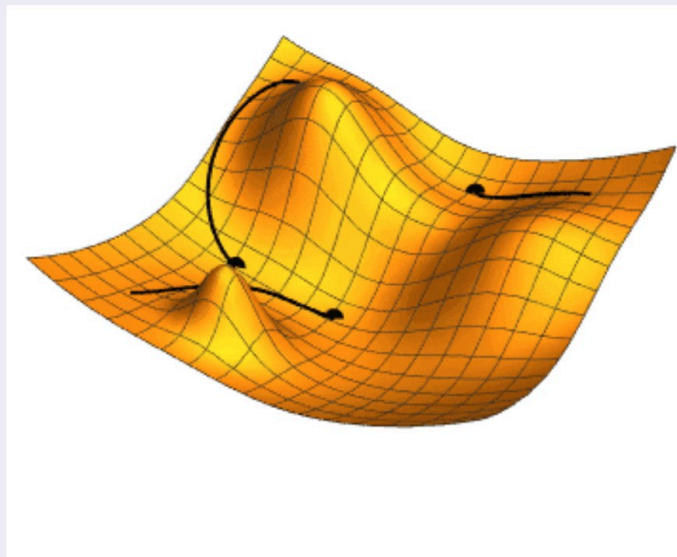


image from Wikimedia



Any questions?

