Fitness Landscape

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Recap

- We need three key elements for SBSE
 - Representation: how we express candidate solutions for storage
 - Fitness Function: how we compare candidate solutions for selection
 - Operators: how we modify candidate solutions for trial-and-error

Fitness Landscape

- A spatial view of the search: there is no guarantee that the **actual** optimisation you are working on can be easily visualised spatially. However, this visual analogy is a useful tool when discussing the distribution of the fitness across possible solutions.
- Given a solution space S (a hyperplane), and a fitness function F, a fitness landscape is a hyper dimensional surface that represents F: S→R

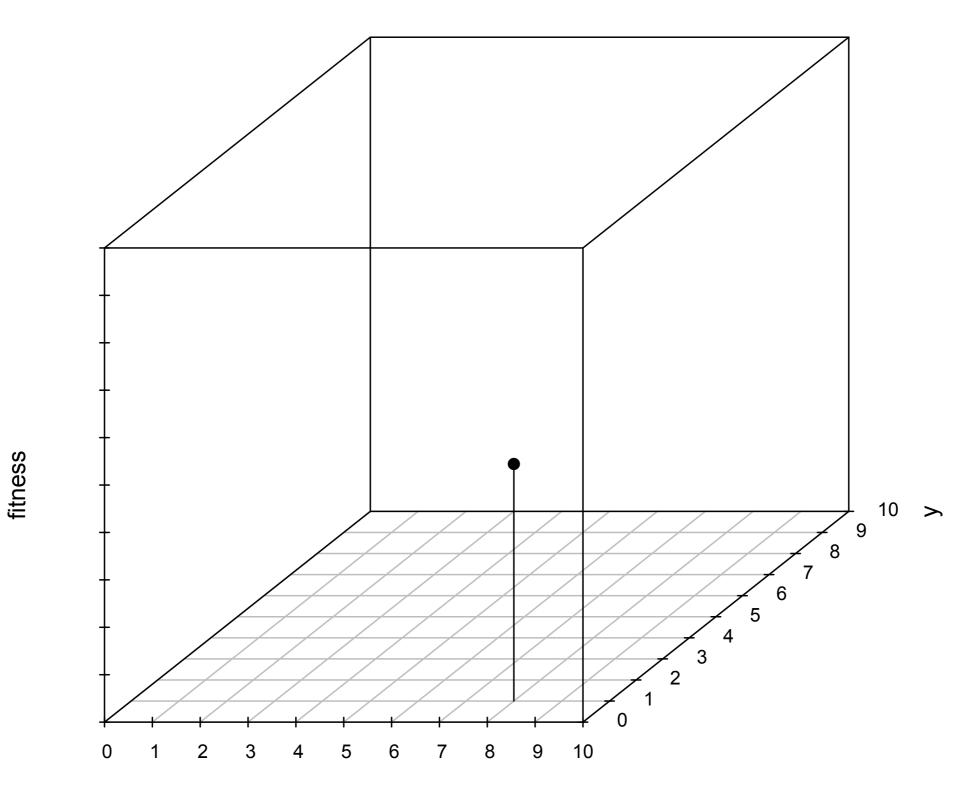
Fitness Landscape

- Let's use a fake problem:
 - Given $0 \le x \le 10, 0 \le y \le 10$, find (x, y)such that x + y = 10.

Solution Space

								0 6				0 0 1 10
		0	0	0	0	0	0	0	0	0	0	0
	∾ –	0	0	0	0	0	0	0	0	0	0	
		0	0	0	0	0	0	0	0	0	0	0
	4 -	0	0	0	0	0	0	0	0	0	0	0
Х		0	0	0	0	0	0	0	0	0	0	0
	9 –	0	0	0	0	0	0	0	0	0	0	0
		0	0	0	0	0	0	0	0	0	0	0
	∞ –	0	0	0	0	0	0	0	0	0	0	0
		0	0	0	0	0	0	0	0	0	0	0
	6 -	0	0	0	0	0	0	0	0	0	0	0

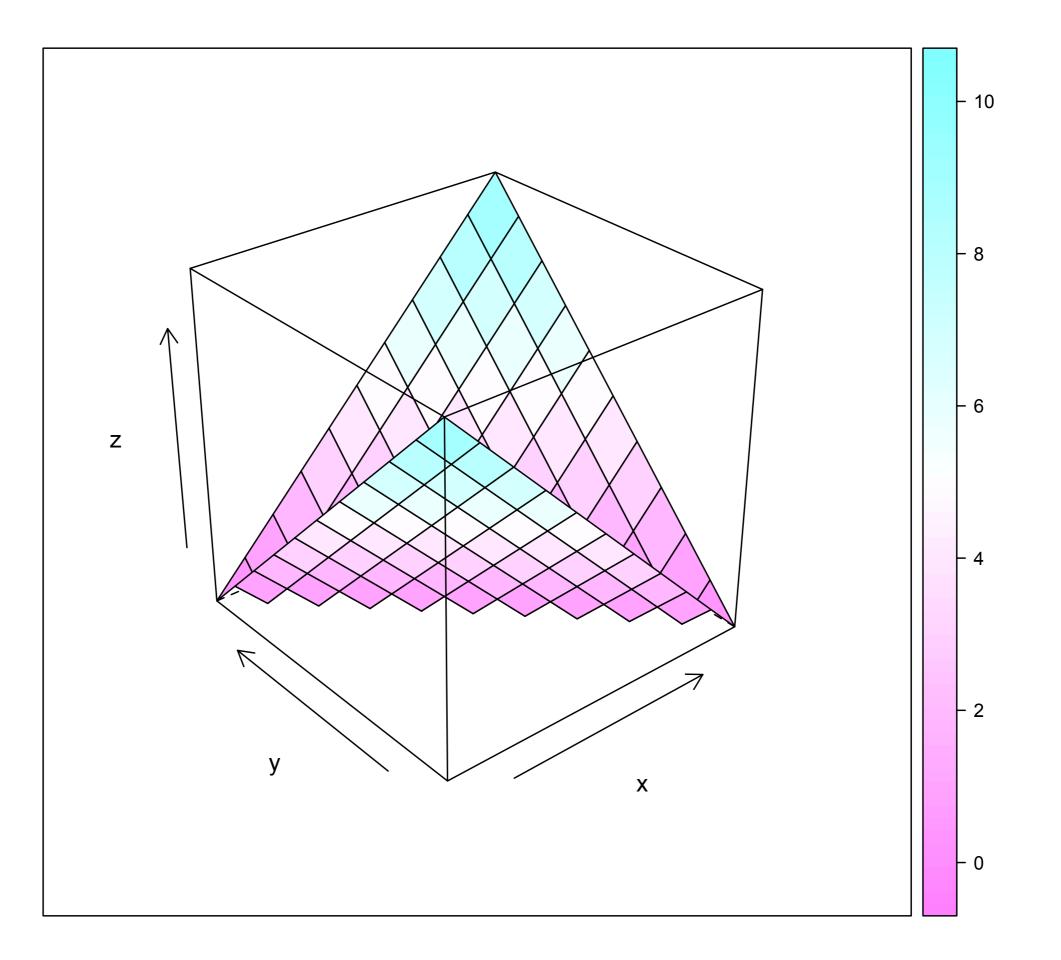
A single point in fitness landscape



Х

Fitness for Fake Problem

- Given (x, y), how far are we from solving the problem?
 - We solve the problem when x + y == 10
 - If the current sum of x and y are s, we are |10 s | far away from solving the solution
 - f(x, y) = |10 (x + y)|
 - Minimise the above function until it becomes 0.

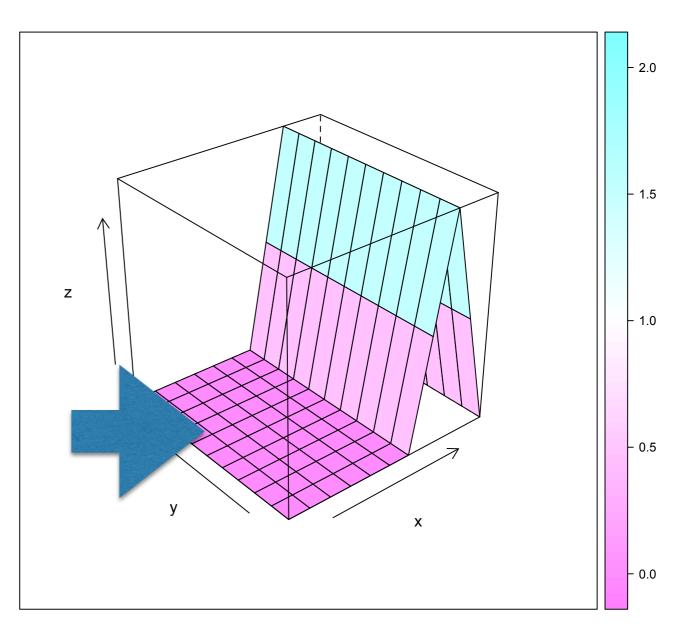


Properties of Landscape

- Size: small/large but also finite/(effectively) infinite
- Flatness: is there a large plateau?
- Ruggedness: how many local optima should we expect?
- Discreteness: continuous numeric, discrete numeric, combinatoric

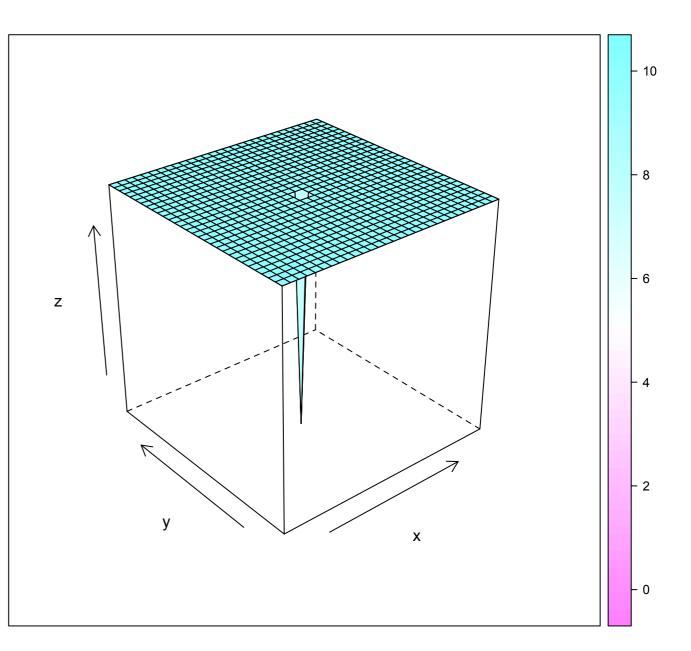
Plateau

- Large, flat region that does not exhibit any gradient.
- Suppose current solution as well as others generated by operators all fall in a plateau.
- There is no guidance; hard to escape.



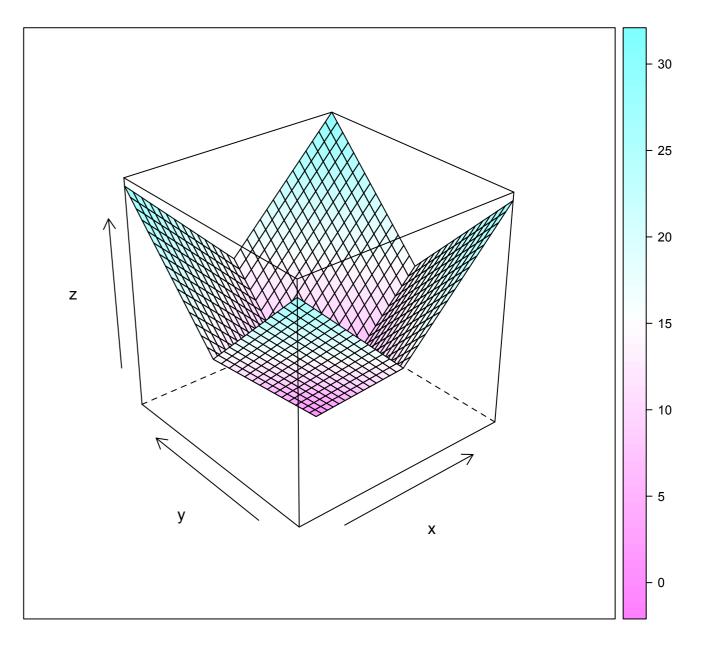
Needle in the Haystack

- Worst landscape to search.
- Can be avoided by transforming the problem and/ or designing better fitness functions
- To search for (x, y) = (15, 15):
 - f1(x, y) = (x==15 && y == 15) ? 0 : 10



Needle in the Haystack

• f2(x, y) = |x-15| + |y-15|



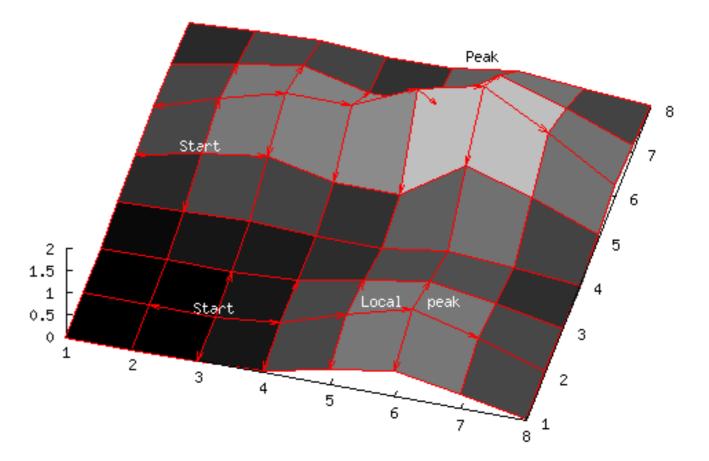
(..later application in testing)

```
if(x == 42){
    //do some computation
    //that needs to be tested
}
if(flag){
    //do some computation
    //that needs to be tested
}
```

M. Harman, L. Hu, R. Hierons, J. Wegener, H. Sthamer, A. Baresel, and M. Roper. Testability trans- formation. IEEE Transactions on Software Engineering, 30(1):3–16, Jan. 2004.

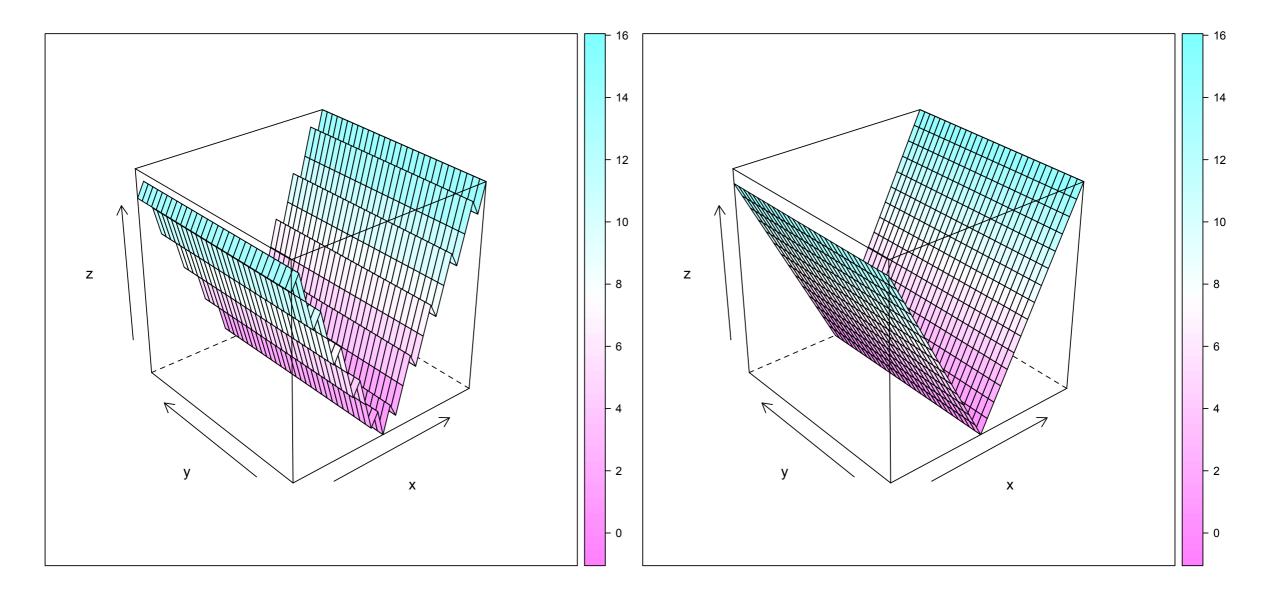
Local vs. Global Optima

- Local optima: better fitness than any surrounding region, but not the best possible fitness
- Global optima: better fitness than any other point in the landscape



Foundation of Genetic Programmin

Ruggedness



Easy to get stuck in one of many local optima

Smooth descent

Discrete Fitness Landscape

- In case of (x, y) = (15, 15), it is (relatively) obvious what the neighbouring solutions are.
 - (14, 15), (16, 15), (15, 14), (15, 16)
 - (16, 16), (14, 14), (16, 14), (14, 16)
- What if we are searching for non-numeric solution?
 - Set membership (e.g. Do I include this requirement or not?, Do I execute this test case or not?)
 - Permutations (e.g. In which order should I execute this test suite?)
 - Highly structured data (e.g. To test this compiler, which program should I use as input?)