Playground

Abdelrahman

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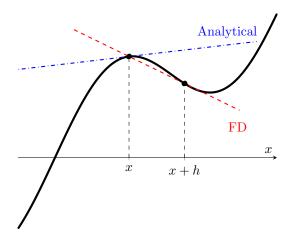


Figure 1: Forward Finite Difference

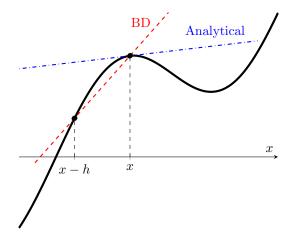


Figure 2: Backward Finite Difference

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Algorithm 1: Hill Climbing Algorithm
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 \begin{aligned} \mathbf{Data:} \ x_i \\ \mathbf{Result:} \ x_n &\leftarrow \text{last solution candidate} \\ x_0 &\leftarrow \text{starting solution candidate}; \\ \mathbf{for} \ i \in \{0, \dots, n\} \ \mathbf{do} \\ & x_{\text{neighbor}} \leftarrow \text{select best neighbor from } x_i \text{ neighborhood}; \\ & \Delta \cos t \leftarrow \text{compute cost difference between } x_{\text{neighbor}} \text{ and } x_i; \\ & \mathbf{if} \ \cos(x_{\text{neighbor}}) \text{ is better than } \cos(x_i) \ \mathbf{then} \\ & & Accept \ x_{i+1} \leftarrow x_i; \\ & \mathbf{else} \ \mathbf{if} \ \cos(x_{\text{neighbor}}) \text{ is worse than } \cos(x_i) \ \mathbf{then} \\ & & & \mathbf{Terminate} \ \text{Search}; \end{aligned}
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Algorithm 2: Simulated Annealing Algorithm

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Data: T_i, x_i
Result: x_n \leftarrow \text{last solution candidate}
T_0 \leftarrow \text{initial temperature of cooling schedule};
x_0 \leftarrow starting solution candidate;
for i \in \{0, ..., n\} do
    x_{\text{neighbor}} \leftarrow \text{randomly sample a neighbor from } x_i \text{ neighborhood}
                    using uniform distribution;
    \Delta \cos t \leftarrow \text{compute cost difference between } x_{\text{neighbor}} \text{ and } x_i;
    if cost(x_{neighbor}) is better than cost(x_i) then
         Accept x_{i+1} \leftarrow x_i;
    else if cost(x_{neighbor}) is worse than cost(x_i) then
         if probability e^{-\Delta cost/T_i} then
              Accept x_{i+1} \leftarrow x_i;
         else
              Reject Do nothing;
         end
end
```

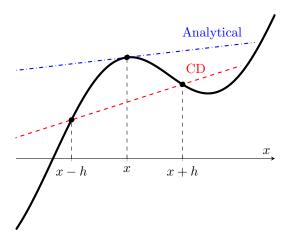


Figure 3: Central Finite Difference

Assumptions		
Steady & Fully developed $\Rightarrow 1$	$\frac{\partial}{\partial t} = 0$, $\frac{\partial}{\partial x} = 0$	
No-Slip Condition $\Rightarrow 2$	$\mathbf{u}(x,0) = 0$, $\mathbf{u}(x,d) = \begin{bmatrix} U \\ 0 \end{bmatrix}$	
Constant Pressure $\Rightarrow 3$	$\frac{\partial p}{\partial x} = 0 \;, \frac{\partial p}{\partial y} = 0$	
Newtonian Fluid	constant viscosity (ν)	
Incompressible	constant density (ρ)	
Laminar & Purely axial	v = 0	

Assumptions	
Steady & Fully developed $\Rightarrow 1$	
No-Slip Condition $\Rightarrow 2$	$\mathbf{u}(x,0) = 0 , \mathbf{u}(x,d) = 0$
Constant Pressure $\Rightarrow 3$	$\frac{\partial p}{\partial x} = -G$, $\frac{\partial p}{\partial y} = 0$
Newtonian Fluid	constant viscosity (ν)
Incompressible	constant density (ρ)
Laminar & Purely axial	v = 0

Assumptions	
Steady & Fully developed $\Rightarrow 1$	$\frac{\partial}{\partial t} = 0$, $\frac{\partial}{\partial x} = 0$
No-Slip Condition $\Rightarrow 2$	$\mathbf{u}(x,y,z) = 0 \iff \sqrt{y^2 + z^2} = R$
Constant Pressure $\Rightarrow 3$	$\frac{\partial p}{\partial x} = -G$, $\frac{\partial p}{\partial y} = 0$, $\frac{\partial p}{\partial z} = 0$
Newtonian Fluid	constant viscosity (ν)
Incompressible	constant density (ρ)
Laminar & Purely axial	v=0, w=0

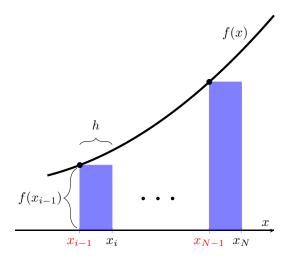


Figure 4: Rectangle Method Left Point Rule

References

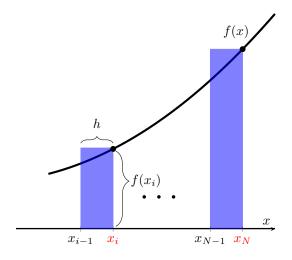


Figure 5: Rectangle Method Right Point Rule

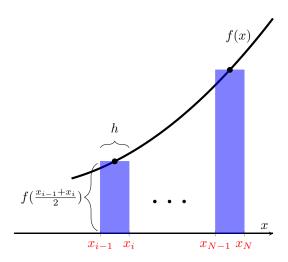


Figure 6: Rectangle Method Midpoint Rule

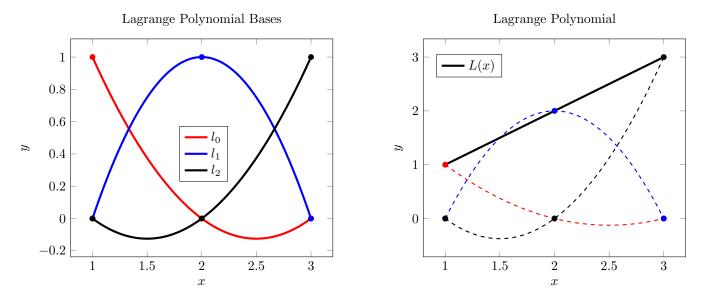


Figure 7: Lagrange Polynomial and its Bases

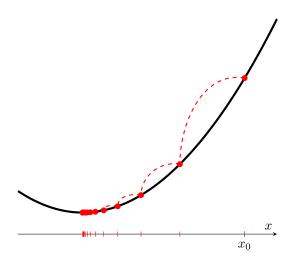


Figure 8: Gradient Decent Steps

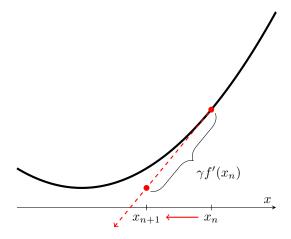


Figure 9: Gradient Decent Step

Neighbor Selection at Iteration (i)

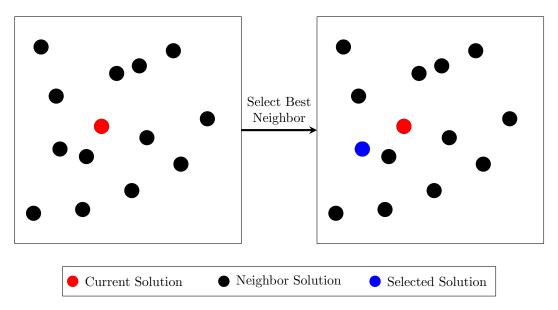


Figure 10: Hill Climbing Neighbor Selection

Neighbor Selection at Iteration (i)

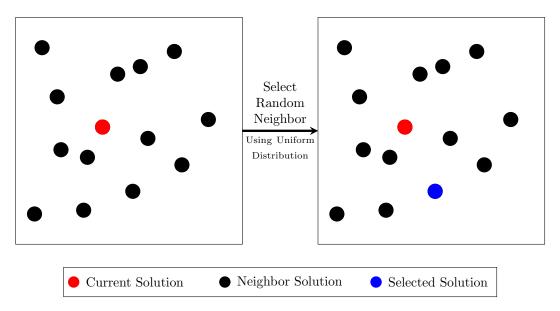


Figure 11: Simulated Annealing Neighbor Selection

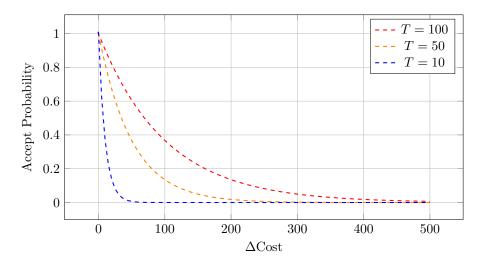


Figure 12: Simulated Annealing Temperature

Evolutionary Process at Generation (i)

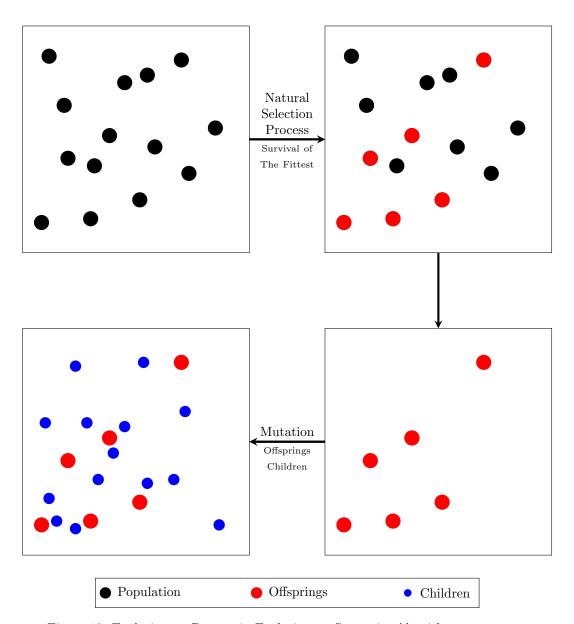


Figure 13: Evolutionary Process in Evolutionary Strategies Algorithm

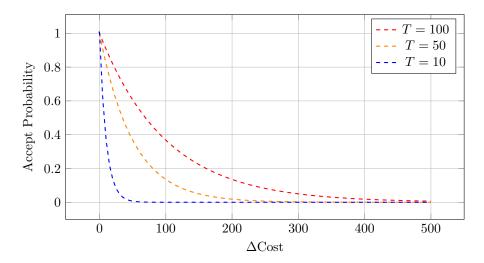


Figure 14: Simulated Annealing Temperature

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Algorithm 3: Simulated Annealing Algorithm
  Data: T_i, x_i
 Result: x_n \leftarrow \text{last solution candidate}
 T_0 \leftarrow \text{initial temperature of cooling schedule};
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 for i \in \{0, ..., n\} do
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      else if cost(x_{neighbor}) is worse than cost(x_i) then
          if probability e^{-\Delta cost/T_i} then
               Accept x_{i+1} \leftarrow x_i;
           else
               Reject Do nothing;
           end
  end
```

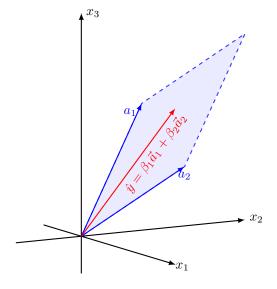


Figure 15: Least Square Problem - Linear combination of column vectors $% \left(1\right) =\left(1\right) +\left(1\right) +\left($

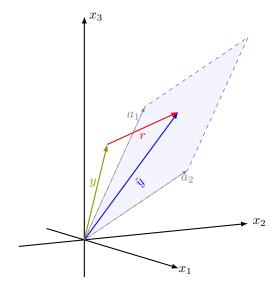


Figure 16: Least Square Problem

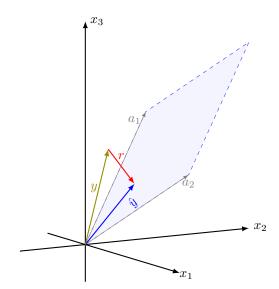


Figure 17: Least Square Problem