

# Population-based algorithms

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ME 564/SYS 564

Wed Oct 17, 2018

Steven Hoffenson



Goal of Week 8: To do another KKT example, and then learn some increasingly-common algorithms that use multiple points in each iteration

# Recap: How to optimize

## 1. **Formulate** the problem

(Weeks 1-2, 4, 9-12)

- a) Define system boundaries
- b) Develop analytical models
- c) Explore/reduce the problem space
- d) Formalize optimization problem

$$\begin{array}{ll}\text{minimize}_{\mathbf{x}} & f(\mathbf{x}, \mathbf{p}) \\ \text{subject to} & \mathbf{g}(\mathbf{x}, \mathbf{p}) \leq 0 \\ & \mathbf{h}(\mathbf{x}, \mathbf{p}) = 0\end{array}$$

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## 2. **Solve** the problem

(Weeks 3, 5-8, 12)

- a) Choose the right approach/algorithm
- b) Solve (by hand, code, or software)
- c) Interpret the results
- d) Iterate if needed

$$\mathbf{x}_{k+1} = \mathbf{x}_k - [\mathbf{H}(\mathbf{x}_k)]^{-1} \nabla f(\mathbf{x}_0)$$

TODAY

# Recap: Gradient-based methods

- **Unconstrained**

- **FONC and SOSC** (when math is simple enough for algebra)
- **Gradient descent** (algorithm with linear convergence)
- **Newton method** (algorithm with super-linear convergence)

- **Constrained**

- **Reduced gradient** (with known active constraints)
- **Generalized Reduced Gradient** (algorithm w active constr)
- **Active set strategy** (algorithm w updating set of active constr)
- **Lagrangian** (equality or active inequality constr)
- **KKT conditions** (with any inequality and equality constr)
- **Quasi-Newton methods** (2<sup>nd</sup>-derivative-free)
- **SQP** (efficiently handles constraints)

# Recap: Quasi-Newton and SQP

**Quasi-Newton methods** approximate  $\mathbf{H}^{-1}$  to simplify math

1. Begin with  $\mathbf{x}_0$  and some assumed  $\mathbf{H}_0^{-1}$ .
2. For iteration  $k$ , set  $\mathbf{x}_{k+1} = \mathbf{x}_k - \alpha_k \mathbf{H}_k^{-1} \nabla f(\mathbf{x}_k)$ .
3. Compute an update matrix  $\mathbf{H}_k^{-1}$  as some function of  $[\nabla f(\mathbf{x}_{k+1}) - \nabla f(\mathbf{x}_k)]$ ,  $[\mathbf{x}_{k+1} - \mathbf{x}_k]$ , and  $\mathbf{H}_k^{-1}$ .
4. Update inverse Hessian approximation:  $\mathbf{H}_{k+1}^{-1} = \mathbf{H}_k^{-1} + \hat{\mathbf{H}}_k^{-1}$ .

**Sequential Quadratic Programming (SQP)** algorithm solves a subproblem for the step size and direction  $\mathbf{s}_k$ , then moves as  $\mathbf{x}_{k+1} = \mathbf{x}_k + \mathbf{s}_k$

$$\begin{aligned} \underset{\mathbf{s}_k}{\text{minimize}} \quad & q(\mathbf{s}_k) = f(\mathbf{x}_k) + \nabla f(\mathbf{x}_k)^T \mathbf{s}_k + \frac{1}{2} \mathbf{s}_k^T \nabla_{xx}^2 \mathcal{L}(\mathbf{x}_k, \lambda_k, \mu_k) \mathbf{s}_k \\ \text{where} \quad & \mathcal{L}(\mathbf{x}_k, \lambda_k, \mu_k) = f(\mathbf{x}_k) - \lambda^T \mathbf{h}(\mathbf{x}_k) - \mu^T \mathbf{g}(\mathbf{x}_k) \\ \text{subject to} \quad & \mathbf{g}(\mathbf{x}_k) + \nabla \mathbf{g}(\mathbf{x}_k)^T \mathbf{s}_k \leq \mathbf{0} \\ & \mathbf{h}(\mathbf{x}_k) + \nabla \mathbf{h}(\mathbf{x}_k)^T \mathbf{s}_k = \mathbf{0} \end{aligned}$$

# Example: KKT conditions (5.14)

$$\min f(\mathbf{x}) = x_1^2 + x_2^2 + x_3^2 + 40x_1 + 20x_2 - 3000$$

$$\text{s.t. } g_1(\mathbf{x}) = x_1 - 50 \geq 0$$

$$g_2(\mathbf{x}) = x_1 + x_2 - 100 \geq 0$$

$$g_3(\mathbf{x}) = x_1 + x_2 + x_3 - 150 \geq 0$$

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Recall the KKT conditions:

$$1. \nabla_{\mathbf{x}} \mathcal{L} = \nabla f(\mathbf{x}^*) + \lambda^T \nabla \mathbf{h}(\mathbf{x}^*) + \mu^T \nabla \mathbf{g}(\mathbf{x}^*) = \mathbf{0}^T$$

$$2. \mathbf{h}(\mathbf{x}^*) = \mathbf{0}, \mathbf{g}(\mathbf{x}^*) \leq \mathbf{0}$$

$$3. \lambda \neq \mathbf{0}, \mu \geq \mathbf{0}$$

$$4. \mu^T \mathbf{g}(\mathbf{x}^*) = \mathbf{0}^T$$

# Example: KKT conditions (5.14)

$$\min f(\mathbf{x}) = x_1^2 + x_2^2 + x_3^2 + 40x_1 + 20x_2 - 3000$$

$$\text{s.t. } g_1(\mathbf{x}) = x_1 - 50 \geq 0$$

$$g_2(\mathbf{x}) = x_1 + x_2 - 100 \geq 0$$

$$g_3(\mathbf{x}) = x_1 + x_2 + x_3 - 150 \geq 0$$

The constraints all need to be multiplied by -1 to become negative null!

$$\nabla_{\mathbf{x}} \mathcal{L} = \nabla f(\mathbf{x}^*) + \lambda^T \mathbf{h}(\mathbf{x}^*) + \mu^T \nabla \mathbf{g}(\mathbf{x}^*) = \mathbf{0}^T$$

$$\nabla f(\mathbf{x}^*) = \begin{bmatrix} \partial f / \partial x_1 \\ \partial f / \partial x_2 \\ \partial f / \partial x_3 \end{bmatrix}$$

$$\nabla \mathbf{g}(\mathbf{x}^*) = \begin{bmatrix} \partial g_1 / \partial x_1 & \partial g_1 / \partial x_2 & \partial g_1 / \partial x_3 \\ \partial g_2 / \partial x_1 & \partial g_2 / \partial x_2 & \partial g_2 / \partial x_3 \\ \partial g_3 / \partial x_1 & \partial g_3 / \partial x_2 & \partial g_3 / \partial x_3 \end{bmatrix}$$

$$\nabla_{\mathbf{x}} \mathcal{L} = \begin{bmatrix} 2x_1 + 40 \\ 2x_2 + 20 \\ 2x_3 \end{bmatrix} + \mathbf{0}^T + [\mu_1 \quad \mu_2 \quad \mu_3] \begin{bmatrix} -1 & 0 & 0 \\ -1 & -1 & 0 \\ -1 & -1 & -1 \end{bmatrix} = \mathbf{0}^T$$

# Example: KKT conditions (5.14)

$$\nabla_{\mathbf{x}} \mathcal{L} = \begin{bmatrix} 2x_1 + 40 \\ 2x_2 + 20 \\ 2x_3 \end{bmatrix} + \mathbf{0}^T + [\mu_1 \quad \mu_2 \quad \mu_3] \begin{bmatrix} -1 & 0 & 0 \\ -1 & -1 & 0 \\ -1 & -1 & -1 \end{bmatrix} = \mathbf{0}^T$$

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Recall the KKT conditions:

1.  $\nabla_{\mathbf{x}} \mathcal{L} = \mathbf{0}^T$
2.  $\mathbf{h}(\mathbf{x}^*) = \mathbf{0}, \mathbf{g}(\mathbf{x}^*) \leq \mathbf{0}$
3.  $\lambda \neq \mathbf{0}, \mu \geq \mathbf{0}$
4.  $\mu^T \mathbf{g}(\mathbf{x}^*) = \mathbf{0}^T$

$$2x_1 + 40 - \mu_1 - \mu_2 - \mu_3 = 0$$

$$2x_2 + 20 - \mu_2 - \mu_3 = 0$$

$$2x_3 - \mu_3 = 0$$

3 equations,  
6 unknowns

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$$\mu^T \mathbf{g}(\mathbf{x}^*) = \mathbf{0}^T$$

$$[\mu_1 \quad \mu_2 \quad \mu_3] \begin{bmatrix} -x_1 + 50 \\ -x_1 - x_2 + 100 \\ -x_1 - x_2 - x_3 + 150 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

3 more equations!

# Example: KKT conditions (5.14)

$$2x_1 + 40 - \mu_1 - \mu_2 - \mu_3 = 0$$

$$2x_2 + 20 - \mu_2 - \mu_3 = 0$$

$$2x_3 - \mu_3 = 0$$

$$\mu_1(-x_1 + 50) = 0$$

$$\mu_2(-x_1 - x_2 + 100) = 0$$

$$\mu_3(-x_1 - x_2 - x_3 + 150) = 0$$

6 equations,  
6 unknowns

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How do we solve?  
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From the lower 3 equations, there are multiple possibilities in each:  
Either  $\mu_i = 0$ , or the expression in parentheses = 0.

So, we will have to examine 8 different scenarios for this system of equations, and check each solution against the remaining KKT conditions:  ~~$\mathbf{h}(\mathbf{x}^*) = \mathbf{0}$~~ ,  $\mathbf{g}(\mathbf{x}^*) \leq \mathbf{0}$ ,  ~~$\boldsymbol{\lambda} \neq \mathbf{0}$~~ ,  $\boldsymbol{\mu} \geq \mathbf{0}$ .

We don't need these, since there are no equality constraints in this problem



# Example: KKT conditions (5.14)

**Scenario 1:**  $\mu_1 = 0, \mu_2 = 0, \mu_3 = 0$

$$2x_1 + 40 - \mu_1 - \mu_2 - \mu_3 = 0$$

$$2x_2 + 20 - \mu_2 - \mu_3 = 0$$

$$2x_3 - \mu_3 = 0$$

$$\mu_1(-x_1 + 50) = 0$$

$$\mu_2(-x_1 - x_2 + 100) = 0$$

$$\mu_3(-x_1 - x_2 - x_3 + 150) = 0$$

$$2x_1 + 40 = 0$$

$$2x_2 + 20 = 0$$

$$2x_3 = 0$$

$$0 = 0$$

$$0 = 0$$

$$0 = 0$$

$$x_1 = -20$$

$$x_2 = -10$$

$$x_3 = 0$$

$$\mu_1 = 0$$

$$\mu_2 = 0$$

$$\mu_3 = 0$$

**Check:**  $g(\mathbf{x}^*) \leq 0, \mu \geq 0$

$$g_1 = 20 + 50 \not\leq 0$$

Since  $g_1$  is violated, this is not a KKT point.

# Example: KKT conditions (5.14)

**Scenario 2:**  $\mu_1 \neq 0, \mu_2 = 0, \mu_3 = 0$

$$2x_1 + 40 - \mu_1 - \mu_2 - \mu_3 = 0$$

$$2x_2 + 20 - \mu_2 - \mu_3 = 0$$

$$2x_3 - \mu_3 = 0$$

$$\mu_1(-x_1 + 50) = 0$$

$$\mu_2(-x_1 - x_2 + 100) = 0$$

$$\mu_3(-x_1 - x_2 - x_3 + 150) = 0$$

$$2x_1 + 40 - \mu_1 = 0$$

$$2x_2 + 20 = 0$$

$$2x_3 = 0$$

$$-x_1 + 50 = 0$$

$$0 = 0$$

$$0 = 0$$

$$x_1 = 50$$

$$x_2 = -10$$

$$x_3 = 0$$

$$\mu_1 = 140$$

$$\mu_2 = 0$$

$$\mu_3 = 0$$

**Check:**  $g(\mathbf{x}^*) \leq 0, \mu \geq 0$

$$g_2 = -50 + 10 + 100 \not\leq 0$$

Since  $g_2$  is violated, this is not a KKT point.

# Example: KKT conditions (5.14)

**Scenario 3:**  $\mu_1 = 0, \mu_2 \neq 0, \mu_3 = 0$

$$2x_1 + 40 - \mu_1 - \mu_2 - \mu_3 = 0$$

$$2x_2 + 20 - \mu_2 - \mu_3 = 0$$

$$2x_3 - \mu_3 = 0$$

$$\mu_1(-x_1 + 50) = 0$$

$$\mu_2(-x_1 - x_2 + 100) = 0$$

$$\mu_3(-x_1 - x_2 - x_3 + 150) = 0$$

$$2x_1 + 40 - \mu_2 = 0$$

$$x_1 = 45$$

$$2x_2 + 20 - \mu_2 = 0$$

$$x_2 = 55$$

$$2x_3 = 0$$

$$x_3 = 0$$

$$0 = 0$$

$$\mu_1 = 0$$

$$-x_1 - x_2 + 100 = 0$$

$$\mu_2 = 130$$

$$0 = 0$$

$$\mu_3 = 0$$

**Check:**  $g(\mathbf{x}^*) \leq 0, \mu \geq 0$

$$g_1 = -45 + 50 \not\leq 0$$

Since  $g_1$  is violated, this is not a KKT point.

# Example: KKT conditions (5.14)

**Scenario 4:**  $\mu_1 = 0, \mu_2 = 0, \mu_3 \neq 0$

$$2x_1 + 40 - \mu_1 - \mu_2 - \mu_3 = 0$$

$$2x_2 + 20 - \mu_2 - \mu_3 = 0$$

$$2x_3 - \mu_3 = 0$$

$$\mu_1(-x_1 + 50) = 0$$

$$\mu_2(-x_1 - x_2 + 100) = 0$$

$$\mu_3(-x_1 - x_2 - x_3 + 150) = 0$$

$$2x_1 + 40 - \mu_3 = 0$$

$$2x_2 + 20 - \mu_3 = 0$$

$$2x_3 - \mu_3 = 0$$

$$0 = 0$$

$$0 = 0$$

$$-x_1 - x_2 - x_3 + 150 = 0$$

$$x_1 = 40$$

$$x_2 = 50$$

$$x_3 = 60$$

$$\mu_1 = 0$$

$$\mu_2 = 0$$

$$\mu_3 = 120$$

**Check:**  $g(\mathbf{x}^*) \leq 0, \mu \geq 0$

$$g_1 = -40 + 50 \not\leq 0$$

Since  $g_1$  is violated, this is not a KKT point.

# Example: KKT conditions (5.14)

**Scenario 5:**  $\mu_1 \neq 0, \mu_2 \neq 0, \mu_3 = 0$

$$2x_1 + 40 - \mu_1 - \mu_2 - \mu_3 = 0$$

$$2x_2 + 20 - \mu_2 - \mu_3 = 0$$

$$2x_3 - \mu_3 = 0$$

$$\mu_1(-x_1 + 50) = 0$$

$$\mu_2(-x_1 - x_2 + 100) = 0$$

$$\mu_3(-x_1 - x_2 - x_3 + 150) = 0$$

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$$2x_1 + 40 - \mu_1 - \mu_2 = 0 \quad x_1 = 50$$

$$2x_2 + 20 - \mu_2 = 0 \quad x_2 = 50$$

$$2x_3 = 0 \quad x_3 = 0$$

$$-x_1 + 50 = 0 \quad \mu_1 = 20$$

$$-x_1 - x_2 + 100 = 0 \quad \mu_2 = 120$$

$$0 = 0 \quad \mu_3 = 0$$

Check:  $g(\mathbf{x}^*) \leq 0, \mu \geq 0$

$$g_3 = -50 - 50 - 0 + 150 \not\leq 0$$

Since  $g_3$  is violated, this is not a KKT point.

# Example: KKT conditions (5.14)

**Scenario 6:**  $\mu_1 \neq 0, \mu_2 = 0, \mu_3 \neq 0$

$$2x_1 + 40 - \mu_1 - \mu_2 - \mu_3 = 0$$

$$2x_2 + 20 - \mu_2 - \mu_3 = 0$$

$$2x_3 - \mu_3 = 0$$

$$\mu_1(-x_1 + 50) = 0$$

$$\mu_2(-x_1 - x_2 + 100) = 0$$

$$\mu_3(-x_1 - x_2 - x_3 + 150) = 0$$

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$$2x_1 + 40 - \mu_1 - \mu_3 = 0$$

$$2x_2 + 20 - \mu_3 = 0$$

$$2x_3 - \mu_3 = 0$$

$$-x_1 + 50 = 0$$

$$0 = 0$$

$$-x_1 - x_2 - x_3 + 150 = 0$$

$$x_1 = 50$$

$$x_2 = 45$$

$$x_3 = 55$$

$$\mu_1 = 30$$

$$\mu_2 = 0$$

$$\mu_3 = 110$$

Check:  $g(x^*) \leq 0, \mu \geq 0$

$$g_2 = -50 - 45 + 100 \not\leq 0$$

Since  $g_2$  is violated, this is not a KKT point.

# Example: KKT conditions (5.14)

**Scenario 7:**  $\mu_1 = 0, \mu_2 \neq 0, \mu_3 \neq 0$

$$2x_1 + 40 - \mu_1 - \mu_2 - \mu_3 = 0$$

$$2x_2 + 20 - \mu_2 - \mu_3 = 0$$

$$2x_3 - \mu_3 = 0$$

$$\mu_1(-x_1 + 50) = 0$$

$$\mu_2(-x_1 - x_2 + 100) = 0$$

$$\mu_3(-x_1 - x_2 - x_3 + 150) = 0$$

$$2x_1 + 40 - \mu_2 - \mu_3 = 0$$

$$2x_2 + 20 - \mu_2 - \mu_3 = 0$$

$$2x_3 - \mu_3 = 0$$

$$0 = 0$$

$$-x_1 - x_2 + 100 = 0$$

$$-x_1 - x_2 - x_3 + 150 = 0$$

$$x_1 = 45$$

$$x_2 = 55$$

$$x_3 = 50$$

$$\mu_1 = 0$$

$$\mu_2 = 30$$

$$\mu_3 = 100$$

**Check:**  $g(\mathbf{x}^*) \leq 0, \mu \geq 0$

$$g_1 = -45 + 50 \not\leq 0$$

Since  $g_1$  is violated, this is not a KKT point.

# Example: KKT conditions (5.14)

**Scenario 8:**  $\mu_1 \neq 0, \mu_2 \neq 0, \mu_3 \neq 0$

$$2x_1 + 40 - \mu_1 - \mu_2 - \mu_3 = 0$$

$$2x_2 + 20 - \mu_2 - \mu_3 = 0$$

$$2x_3 - \mu_3 = 0$$

$$\mu_1(-x_1 + 50) = 0$$

$$\mu_2(-x_1 - x_2 + 100) = 0$$

$$\mu_3(-x_1 - x_2 - x_3 + 150) = 0$$

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$$2x_1 + 40 - \mu_1 - \mu_2 - \mu_3 = 0 \quad x_1 = 50$$

$$2x_2 + 20 - \mu_2 - \mu_3 = 0 \quad x_2 = 50$$

$$2x_3 - \mu_3 = 0 \quad x_3 = 50$$

$$-x_1 + 50 = 0 \quad \mu_1 = 20$$

$$-x_1 - x_2 + 100 = 0 \quad \mu_2 = 20$$

$$-x_1 - x_2 - x_3 + 150 = 0 \quad \mu_3 = 100$$

**Check:**  $g(\mathbf{x}^*) \leq 0, \mu \geq 0$

All of these conditions hold, so  $(50, 50, 50)^T$  is a KKT point!



# Example: KKT conditions (5.14)

$$\min f(\mathbf{x}) = x_1^2 + x_2^2 + x_3^2 + 40x_1 + 20x_2 - 3000$$

$$\text{s.t. } g_1(\mathbf{x}) = x_1 - 50 \geq 0$$

$$g_2(\mathbf{x}) = x_1 + x_2 - 100 \geq 0$$

$$g_3(\mathbf{x}) = x_1 + x_2 + x_3 - 150 \geq 0$$

Across the 8 scenarios, the only KKT point is  $\mathbf{x}^* = (50, 50, 50)^T$ ,  $f^* = 7500$

Now we can test the SOSC:

$$\nabla_{\mathbf{x}} \mathcal{L} = \begin{bmatrix} 2x_1 + 40 - \mu_1 - \mu_2 - \mu_3 \\ 2x_2 + 20 - \mu_2 - \mu_3 \\ 2x_3 - \mu_3 \end{bmatrix}$$

$$\mathcal{L}_{xx} = \begin{bmatrix} 2 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 2 \end{bmatrix} \leftarrow \text{This is pos. def. everywhere!}$$

Therefore,  $\mathbf{x}^*$  is a **global minimizer**.

This is the analytical approach to solving constrained problems. We also have algorithmic approaches like SQP

# Summary: Gradient-based algorithms

- Gradient descent
- Newton method
- Generalized Reduced Gradient (GRG)
- Active set strategy
- Quasi-Newton methods
- Sequential Quadratic Programming (SQP)

$$\nabla f$$

# Gradient-free approaches



- Approximation models
- Pattern search (e.g., Hooke-Jeeves, Nelder-Meade)
- Space-filling search (e.g., DIRECT)
- Random search (e.g., Simulated Annealing)
- Linear Programming (e.g., Simplex)
- Genetic/evolutionary algorithms
- Particle swarm
- Ant colony

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We covered  
some of these  
in Week 3

These are  
population-based  
algorithms

# Approximation models

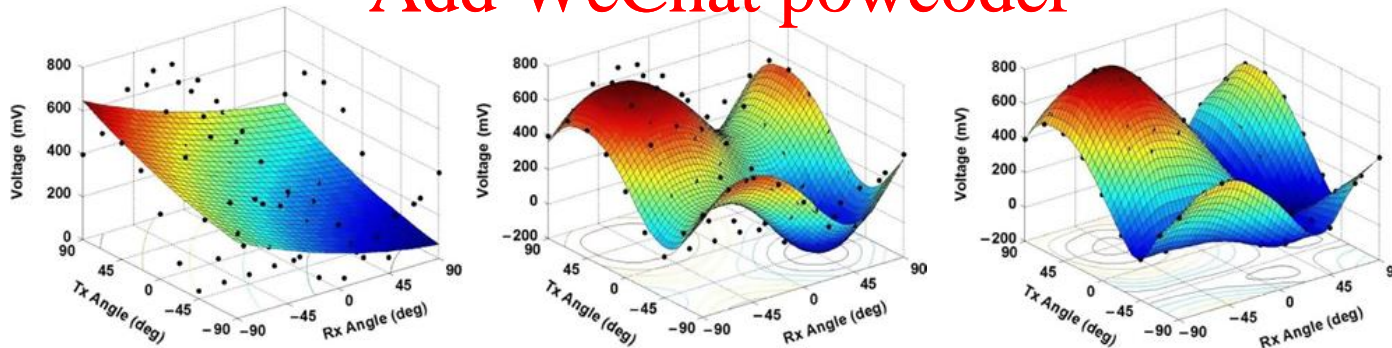
*Use gradient-based methods even though your functions are not differentiable*

1. Approximate derivatives with finite differences

$$f'(x) \approx \frac{f(x+h) - f(x)}{h} \text{ for some small } h$$

2. Metamodels: Sample points & fit a function

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Figures from:

Zamani, A., Mirabadi, A., and Schmid, F. (2012), "Applying metamodeling techniques in the design and optimization of train wheel detector", *Sensor Review*, 32 (4), 327 – 336.

# Population-based algorithms

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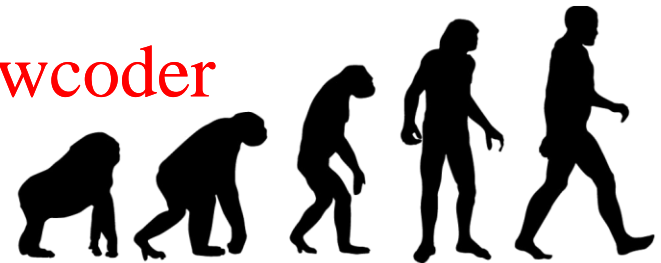
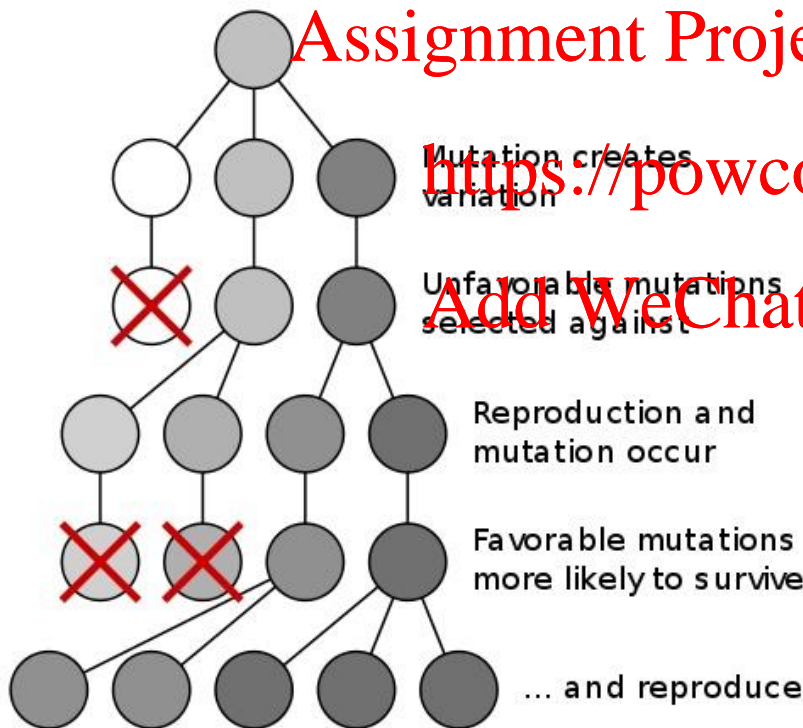
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Goal of Week 8: To learn some increasingly-common algorithms that use multiple points in each iteration, and practice using them

# Genetic algorithms – overview

- A.k.a., evolutionary algorithms
- Mimic gene selection and “survival of the fittest”



# Genetic algorithms – steps

1. Start with a random population (set) of inputs
2. Select the best among population to be “parents”
3. Use parents to spawn a new generation through crossovers and mutations
4. Repeat 2-3 with more “generations” until satisfied

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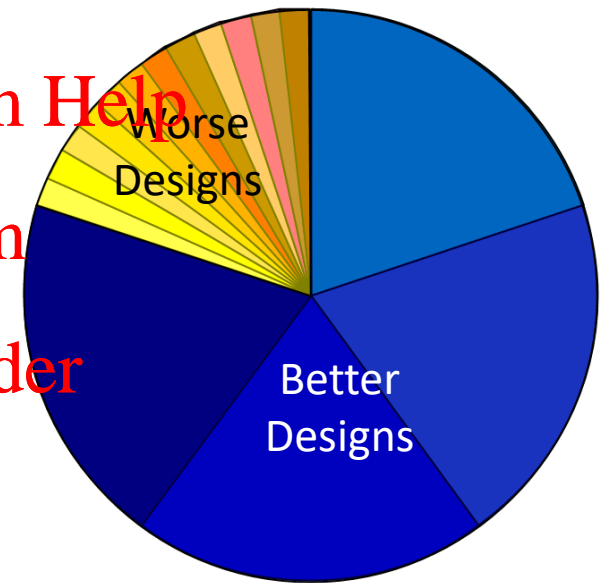
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# Genetic algorithms – parent selection

There are different ways to select parents from a population:

- **Elitism:** pick only the absolute best points
- **Roulette wheel:** better designs have a better chance of being chosen (pictured)
- **Tournament:** segment the population randomly, and choose the best in each segment



Different algorithms use different strategies.



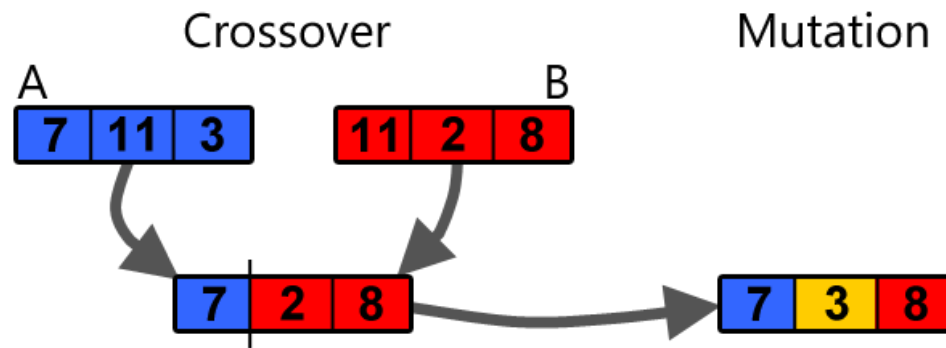
# Genetic algorithms – spawning

Spawning new generations may be done using a combination of:

**1. Survivors:** the best simply join the new generation

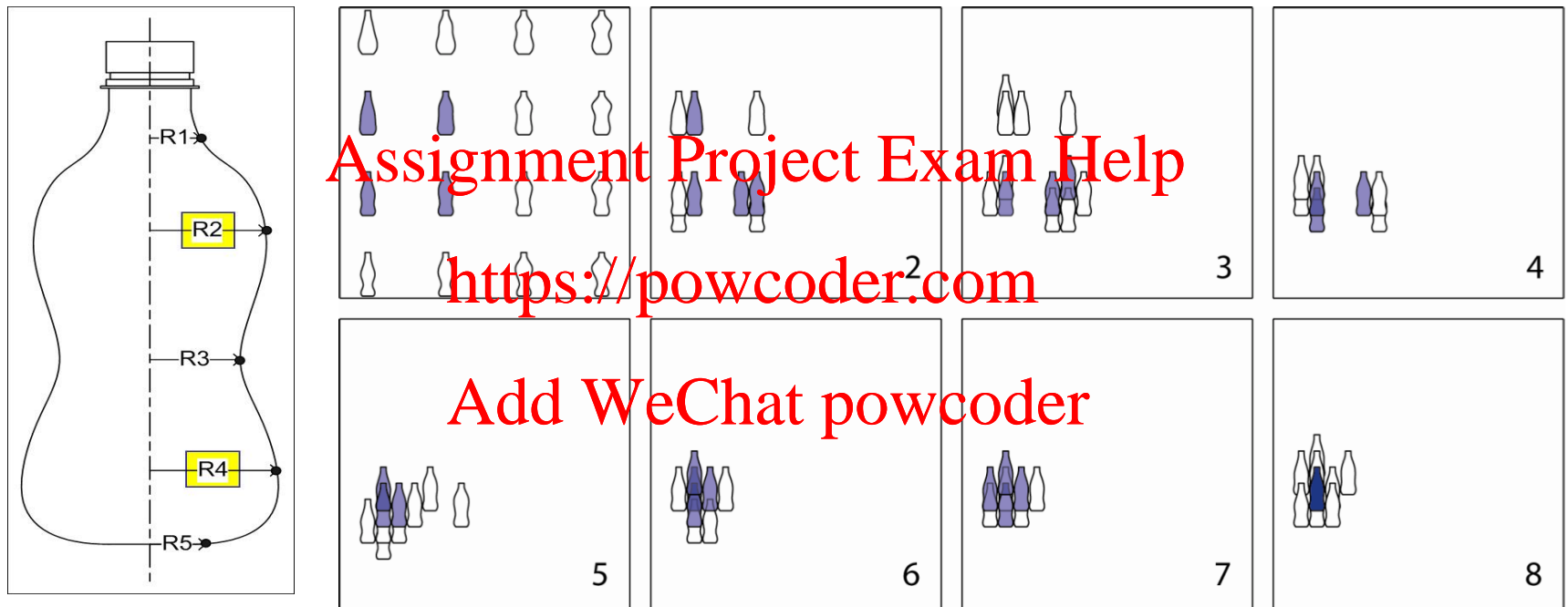
**2. Crossovers:** combinations of traits from 2 parents create a child

**3. Mutations:** traits from a parent randomly change



# Example: *Interactive* Genetic Algorithm

Which bottle shape do you prefer?



The interactive part means that humans do the “parent selection” portion in each iteration of the GA. In this case, it usually converged to a Coca-Cola bottle shape.

# MATLAB – ga

There is a genetic algorithm function 'ga' that can handle objectives and constraints in a similar way to fmincon

[xopt, fopt] = ga(OBJFUN, NVAR, A, b, Aeq, beq, lb, ub, NONLCON)

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Note the difference:

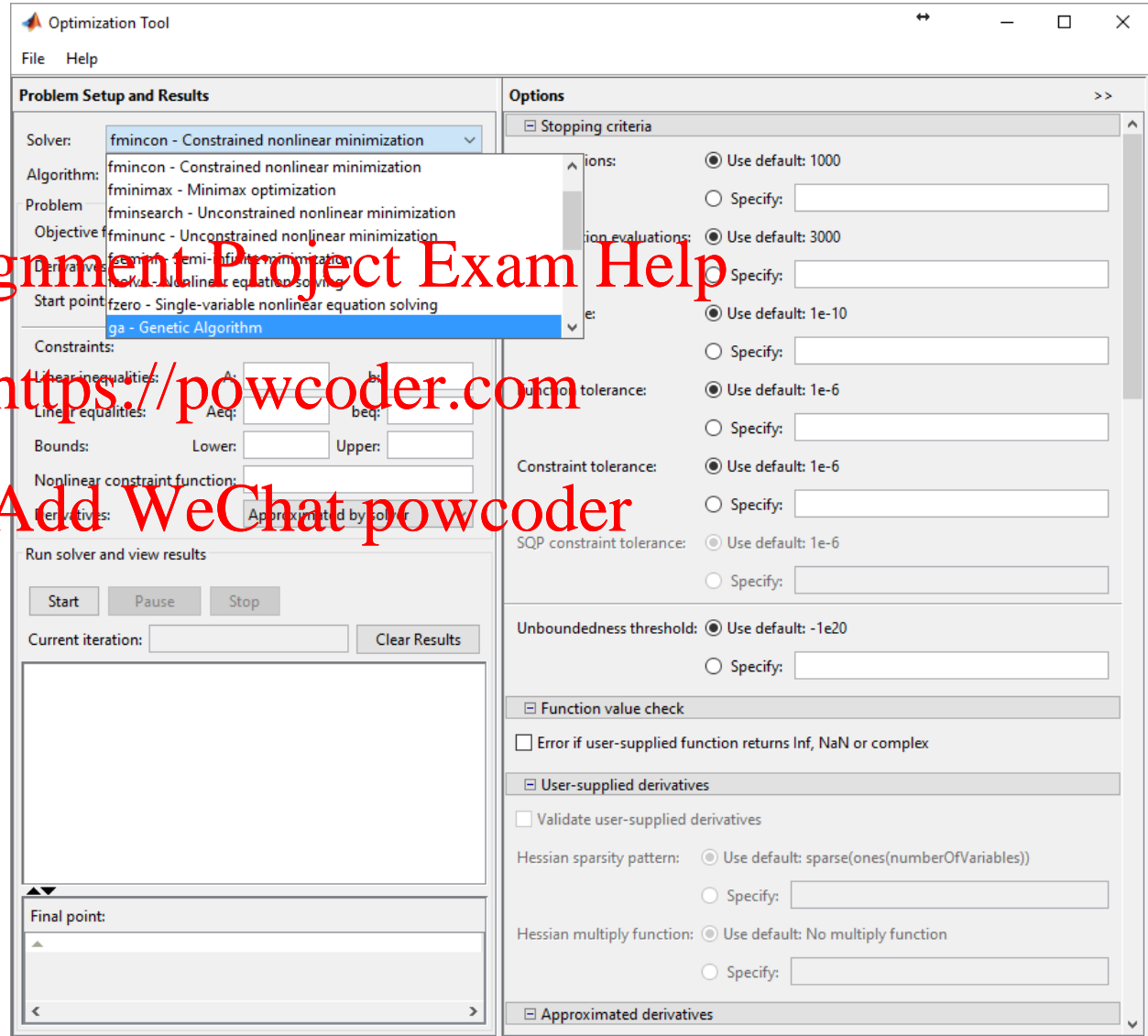
- fmincon needs a start point
- ga just needs the # of variables

**Note:** This function requires the “Global optimization toolbox” to be installed.

# MATLAB – optimtool

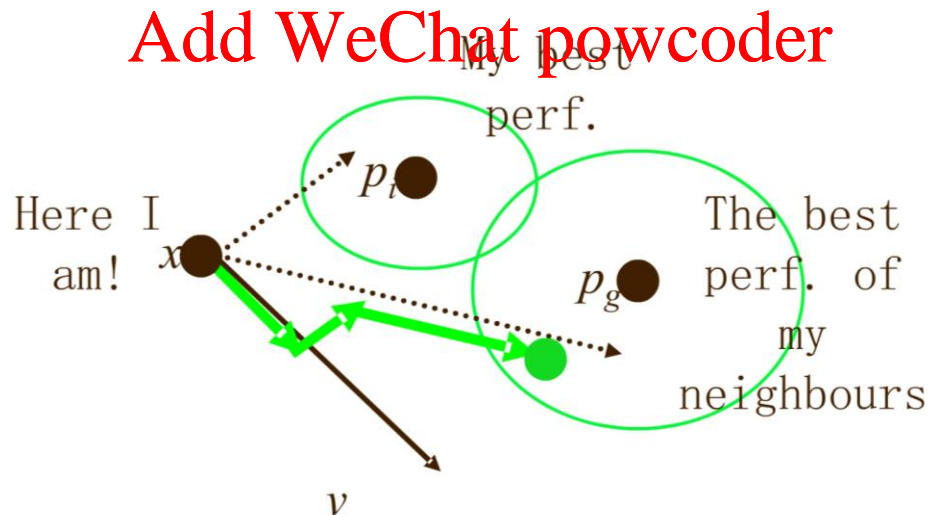
MATLAB currently has an interactive tool 'optimtool' that can guide you through optimizing with different options

Note: This may disappear in future MATLAB releases ☹

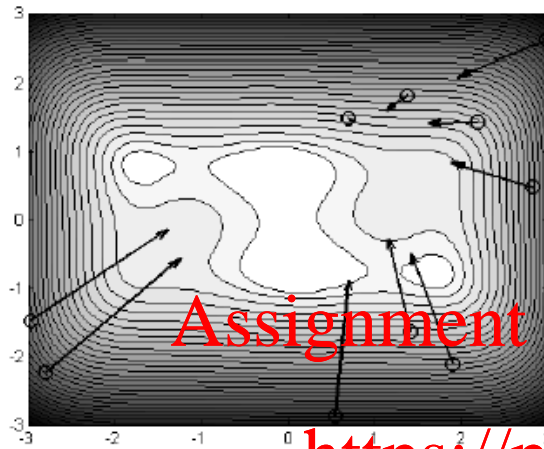


# Particle swarm optimization

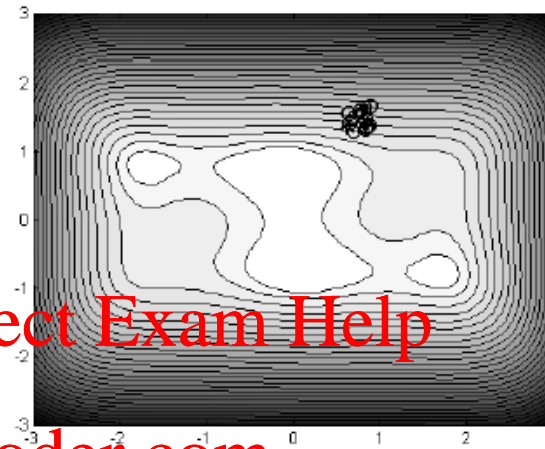
1. Start with some random set of points (particles) with directions/velocities within the input space
2. At each iteration, update each particle's position and velocity based on the particle's best previous position and the best positions of its near neighbors



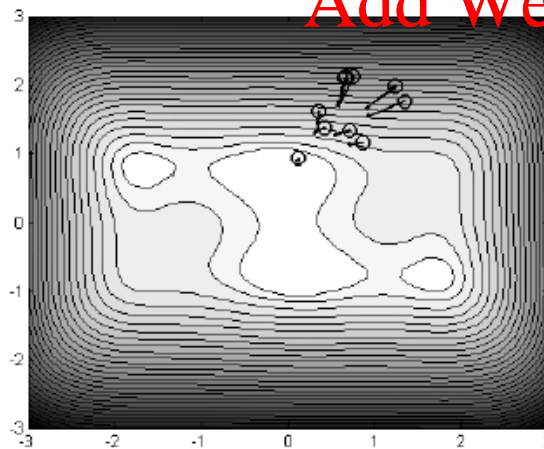
# Particle swarm optimization



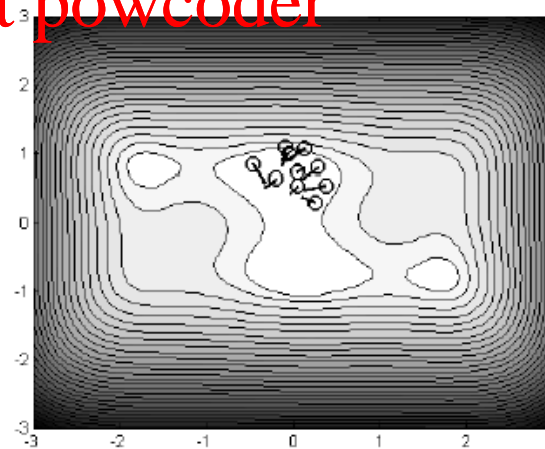
(a) Iteration 1



(b) Iteration 2



(c) Iteration 3



(d) Iteration 4

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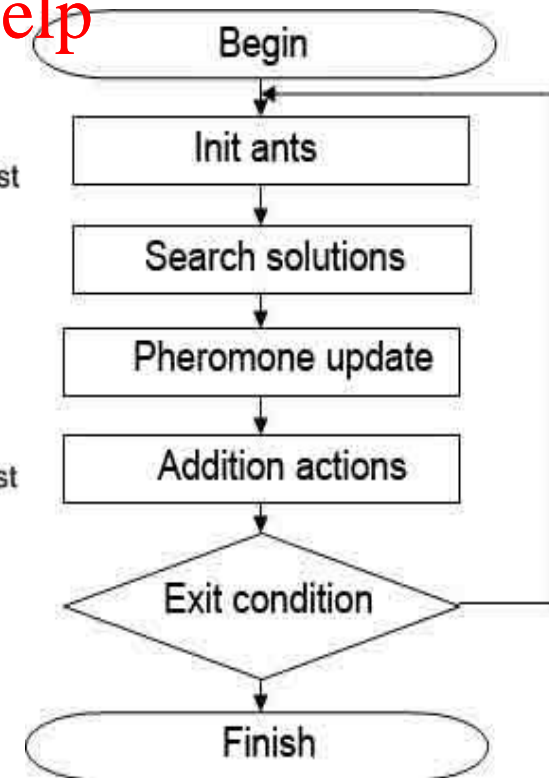
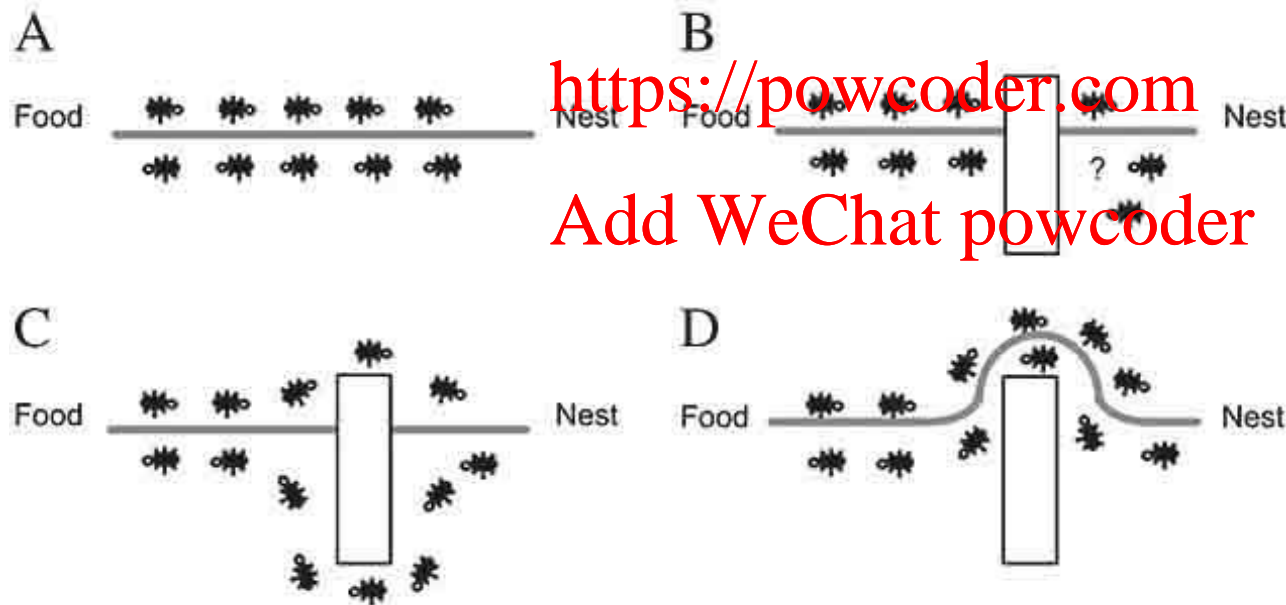
# Ant colony optimization

Adjustments depend on a combination of randomness and pheromones (which tell what has been tried and true), which evaporate over time

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# Non-continuous/discrete variables

What to do when we have **non-continuous** variables?

## Examples:

- Number of rotors on a drone (must be integer)
- Number of cylinders in an engine (must be integer)
- Material choice (there are a few options with different specific properties)

## Strategies:

1. Treat them as continuous, and then pick nearby point
2. Parametric optimization
3. Integer programming



# Treat discrete variables as continuous

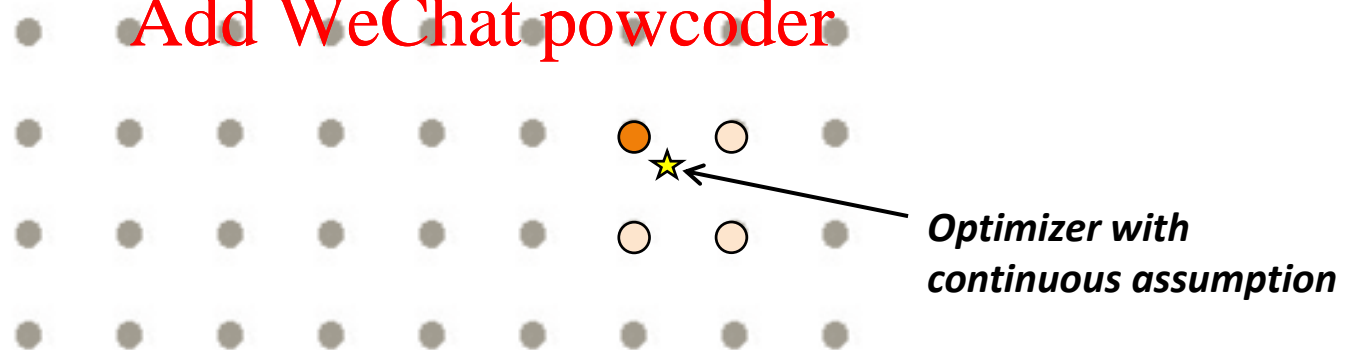
The **easiest** way to handle non-continuous variables (material properties, countable things) is to treat them as continuous:

1. Optimize pretending that they are continuous
2. Choose the discrete value closest to the optimizer (or evaluate all surrounding points and pick the best)

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*Optimizer with  
continuous assumption*

*This often does not work well with non-convex functions!*

# Parametric optimization

When the number of discrete choices is small, you can do ***parametric optimization***:

1. List out all discrete variable combinations
2. Optimize to find the best solution under each of the scenarios in (1)
3. Choose the best from all the solutions in (2)

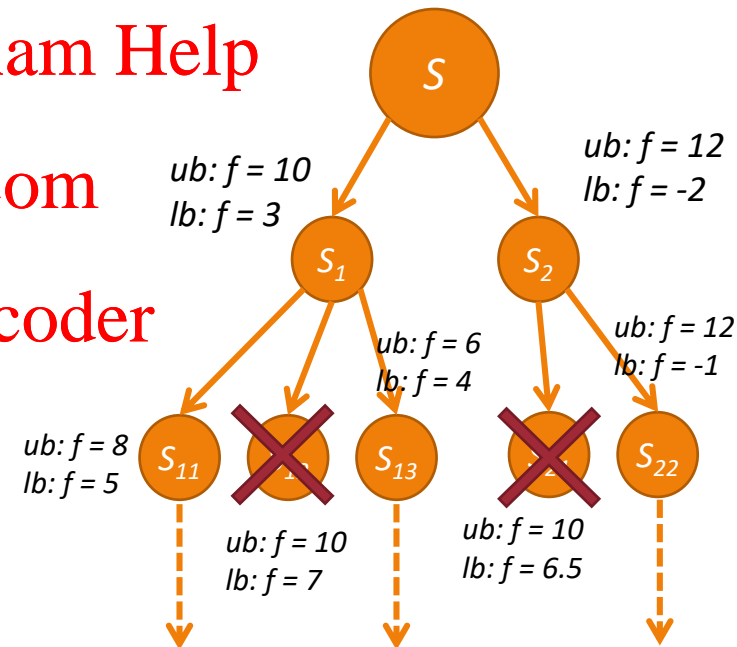
*This works well when there is a small number of discrete choices and the functions are quick to evaluate*

# Discrete/integer programming

*When you have a large number of non-continuous variables...*

## Branch-and-bound algorithm

1. Branch: Split the space of candidate solutions
2. Bound: Compute upper and lower bounds on each subspace (i.e., figure out the highest and lowest possible objective function values for each group)
3. If the lower bound for one group is higher than the upper bound for another, eliminate it
4. Repeat 1-3 until a single solution



# Recap: Gradient-free algorithms

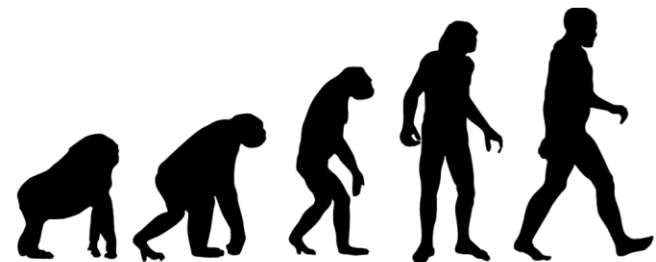
- Approximation models
- Pattern search (e.g., Hooke-Jeeves)
- Space-filling search (e.g., DIRECT)
- Random search (e.g., Simulated Annealing)
- Linear Programming (e.g., Simplex)
- Genetic/evolutionary algorithms
- Particle swarm
- Ant colony

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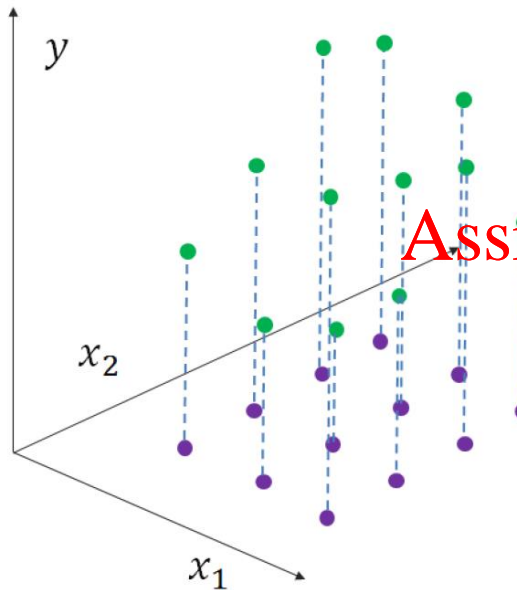
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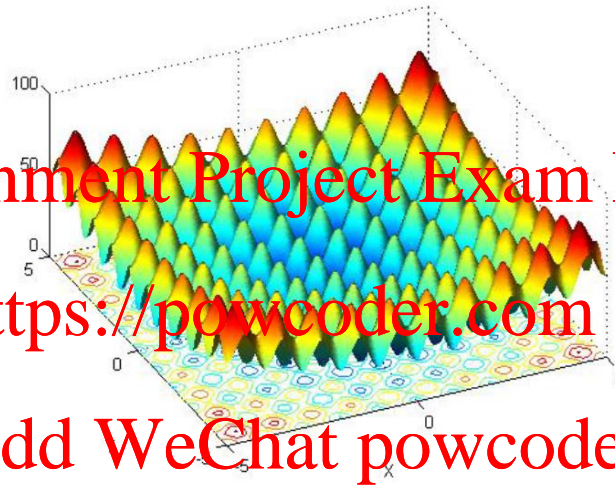
Week 3



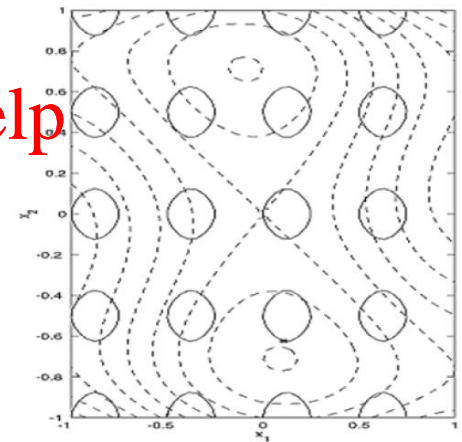
# When to use gradient-free methods



**No Gradient**



**Too Many  
Local Minima**



**Complicated  
Design Space**

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# Cons of gradient-free methods

- Usually slower to converge
  - These require many more function evaluations
  - The search directions are not always efficient
- No optimality guarantee
  - There is no analogy to FONC ( $\text{grad} = 0$ ) and SOSC (Hessian is positive definite)
  - Convergence must be measured by changes to  $f$  and  $x$
- Many parameters to tune
- Constraint handling is imperfect
  - Cannot use Lagrangian
  - Must use penalty or barrier
- Stochastic methods are not repeatable

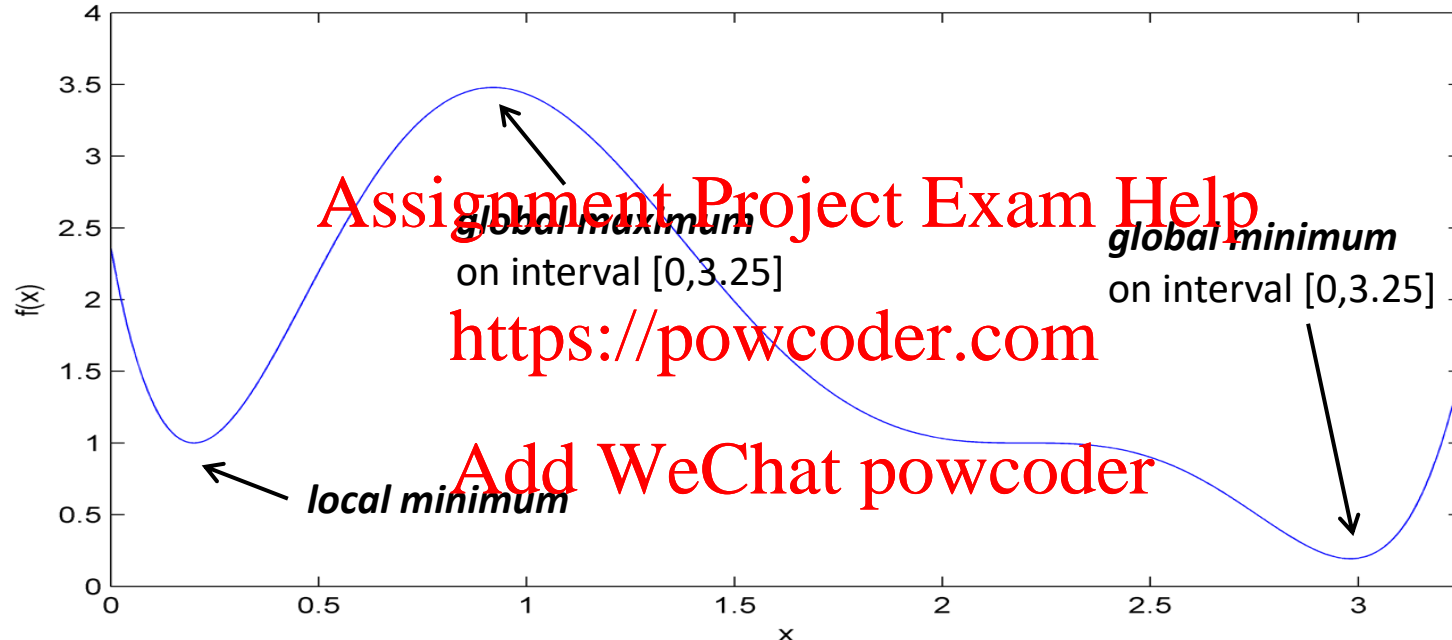
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# A note on global optimization

*Most algorithms seek local optima*



*To find global solutions, try:*

1. Performing optimization with multiple start points
2. Using global algorithms (e.g., genetic algorithms & particle swarm)

# Recap: How to optimize

## 1. **Formulate** the problem

(Weeks 1-2, 4, 9-12)

- a) Define system boundaries
- b) Develop analytical models
- c) Explore/reduce the problem space
- d) Formalize optimization problem

$$\begin{array}{ll}\text{minimize}_{\mathbf{x}} & f(\mathbf{x}, \mathbf{p}) \\ \text{subject to} & \mathbf{g}(\mathbf{x}, \mathbf{p}) \leq 0 \\ & \mathbf{h}(\mathbf{x}, \mathbf{p}) = 0\end{array}$$

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## 2. **Solve** the problem

(Weeks 3, 5-8, 12)

- a) Choose the right approach/algorithm
- b) Solve (by hand, code, or software)
- c) Interpret the results
- d) Iterate if needed

$$\mathbf{x}_{k+1} = \mathbf{x}_k - [\mathbf{H}(\mathbf{x}_k)]^{-1} \nabla f(\mathbf{x}_0)$$

TODAY



# Summary of optimization approaches

- Mathematical/analytical
  - Elimination of constraints using monotonicity analysis
  - Finding stationary points and their nature with optimality conditions
- Linear programming: Simplex
- Nonlinear gradient-based methods
  - Gradient descent
  - Newton method
  - Generalized Reduced Gradient (GRG)
  - Active set strategies
  - Quasi-Newton methods
  - Sequential quadratic programming (SQP)
- Nonlinear gradient-free methods
  - Approximation
  - Pattern search
  - Space-filling search
  - Random search
  - Genetic/evolutionary algorithms (GAs/EAs)
  - Particle swarm & Ant colony
- Integer programming: Branch-and-bound

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# When to use what?

- If the **math is simple** enough, try solving by hand (monotonicity analysis, optimality conditions: FONC/SOSC or KKT)
- If you have a **linear** problem, Simplex LP is efficient
- If the **functions are way too slow** to evaluate, create a meta-model
- If you have a **convex and differentiable** problem, you can't beat the efficiency and accuracy of a gradient-based algorithm (e.g., SQP, GRG)
- If you have a **convex, non-differentiable** problem, pattern-search or random search algorithms are usually effective
- If you have **non-continuous variables**, you should either:
  - Solve parametrically (i.e., solve separately for each discrete value)
  - Use branch-and-bound techniques
- With tricky problems (**non-linear, non-convex**) with fast functions, try:
  - Convex search methods with multiple start points
  - Gradient-free algorithms like GA's and other bio-inspired searches

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# Acknowledgements

- Some of this material came from Chapter 7 of the textbook, *Principles of Optimal Design*
- Some of these slides and examples came from Dr. John Whitefoot, Dr. Alex Burnap, Dr. Yi Ren, and Dr. Michael Kokkolaras at the University of Michigan

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# Announcements

- HW4 is posted, due in 13 days, Tuesday at noon
- Project progress reports are due a week from Sunday
- Please do the mid-semester survey! It's totally anonymous and will help me improve the course

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