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## Unconstrained Minimization Algorithm

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- (1) Initialize  $\mathbf{x}^0$ , set  $k := 0$ .
- (2) **while** *stopping condition is not satisfied at  $\mathbf{x}^k$* 
  - (a) Find  $\mathbf{x}^{k+1}$  such that  $f(\mathbf{x}^{k+1}) < f(\mathbf{x}^k)$ .
  - (b)  $k := k + 1$**endwhile**

**Output :**  $\mathbf{x}^* = \mathbf{x}^k$ , a local minimum of  $f(\mathbf{x})$ .

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- How to find  $\mathbf{x}^{k+1}$  in Step 2(a) of the algorithm?
- Which *stopping condition* can be used?
- Does the algorithm converge? If yes, how fast does it converge?
- Does the convergence and its speed depend on  $\mathbf{x}^0$ ?

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## Unconstrained Minimization Algorithm

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- (1) Initialize  $\mathbf{x}^0$  and  $\epsilon$ , set  $k := 0$ .
- (2) **while**  $\|\mathbf{g}(\mathbf{x}^k)\| > \epsilon$ 
  - (a) Find  $\mathbf{x}^{k+1}$  such that  $f(\mathbf{x}^{k+1}) < f(\mathbf{x}^k)$ .
  - (b)  $k := k + 1$**endwhile**

**Output :**  $\mathbf{x}^* = \mathbf{x}^k$ , a **stationary point** of  $f(\mathbf{x})$ .

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### How to find $\mathbf{x}^{k+1}$ in Step 2(a)?

- Find a *descent direction*  $\mathbf{d}^k$  for  $f$  at  $\mathbf{x}^k$
- Take a step  $\alpha^k (> 0)$  along  $\mathbf{d}^k$  such that
  - $f(\mathbf{x}^{k+1}) < f(\mathbf{x}^k)$
  - $\mathbf{x}^{k+1} = \mathbf{x}^k + \alpha^k \mathbf{d}^k$

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## Unconstrained Minimization Algorithm

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- (1) Initialize  $\mathbf{x}^0$  and  $\epsilon$ , set  $k := 0$ .
- (2) **while**  $\|\mathbf{g}(\mathbf{x}^k)\| > \epsilon$ 
  - (a) Find a descent direction  $\mathbf{d}^k$  for  $f$  at  $\mathbf{x}^k$
  - (b) Find  $\alpha^k (> 0)$  along  $\mathbf{d}^k$  such that  $f(\mathbf{x}^k + \alpha^k \mathbf{d}^k) < f(\mathbf{x}^k)$
  - (c)  $\mathbf{x}^{k+1} = \mathbf{x}^k + \alpha^k \mathbf{d}^k$
  - (d)  $k := k + 1$

**endwhile**

**Output :**  $\mathbf{x}^* = \mathbf{x}^k$ , a stationary point of  $f(\mathbf{x})$ .

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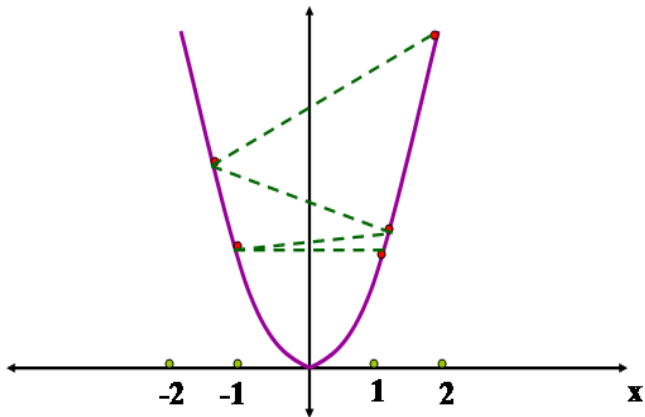
- How to determine  $\alpha^k$  in Step 2(b)?

## Step Length Determination

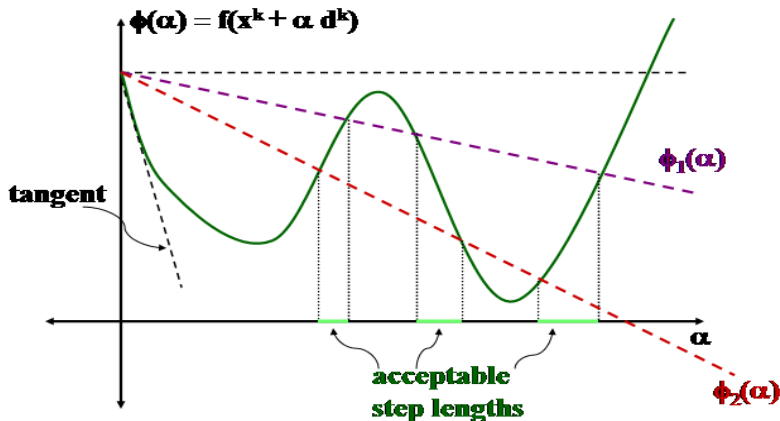
- **Exact Line Search** : Given a descent direction  $\mathbf{d}^k$ , determine  $\alpha^k$  by solving the optimization problem:

$$\alpha^k = \arg \min_{\alpha > 0} \phi(\alpha) \triangleq f(\mathbf{x}^k + \alpha \mathbf{d}^k)$$

- **Inexact Line Search** :
  - Choice of  $\alpha^k$  is crucial



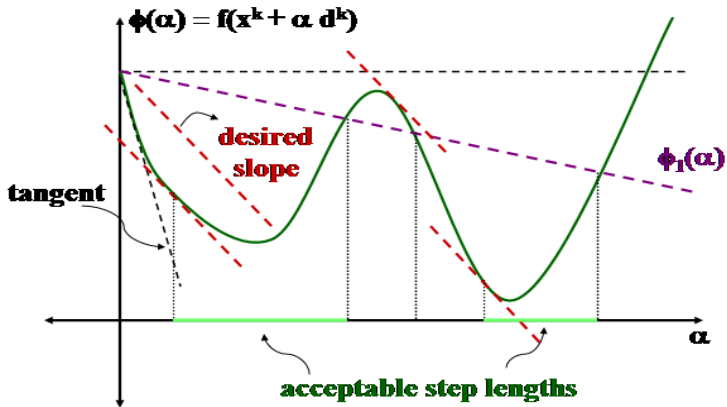
- Small decrease in function values relative to the step length



Armijo-Goldstein Conditions: Choose  $\alpha^k$  such that

$$\phi_2(\alpha^k) \leq f(x^k + \alpha^k d^k) \leq \phi_1(\alpha^k)$$

Wolfe's condition ensures sufficient rate of decrease of function value in the given direction



Choose  $\alpha^k$  such that

$$\phi'(\alpha^k) \geq c_2 \phi'(0), \quad c_2 \in (c_1, 1) \quad \text{Wolfe's Condition}$$