## Descent methods

$$
x^{(k+1)}=x^{(k)}+t^{(k)} \Delta x^{(k)} \text { with } f\left(x^{(k+1)}\right)<f\left(x^{(k)}\right)
$$

- other notations: $x^{+}=x+t \Delta x, x:=x+t \Delta x$
- $\Delta x$ is the step, or search direction; $t$ is the step size, or step length
- from convexity, $f\left(x^{+}\right)<f(x)$ implies $\nabla f(x)^{T} \Delta x<0$ (i.e., $\Delta x$ is a descent direction)

General descent method.
given a starting point $x \in \operatorname{dom} f$. repeat

1. Determine a descent direction $\Delta x$.
2. Line search. Choose a step size $t>0$.
3. Update. $x:=x+t \Delta x$.
until stopping criterion is satisfied.

## Line search types

exact line search: $t=\operatorname{argmin}_{t>0} f(x+t \Delta x)$
backtracking line search (with parameters $\alpha \in(0,1 / 2), \beta \in(0,1)$ )

- starting at $t=1$, repeat $t:=\beta t$ until

$$
f(x+t \Delta x)<f(x)+\alpha t \nabla f(x)^{T} \Delta x
$$

- graphical interpretation: backtrack until $t \leq t_{0}$



## Gradient descent method

general descent method with $\Delta x=-\nabla f(x)$
given a starting point $x \in \operatorname{dom} f$.
repeat

1. $\Delta x:=-\nabla f(x)$.
2. Line search. Choose step size $t$ via exact or backtracking line search.
3. Update. $x:=x+t \Delta x$.
until stopping criterion is satisfied.

- stopping criterion usually of the form $\|\nabla f(x)\|_{2} \leq \epsilon$
- convergence result: for strongly convex $f$,

$$
f\left(x^{(k)}\right)-p^{\star} \leq c^{k}\left(f\left(x^{(0)}\right)-p^{\star}\right)
$$

$c \in(0,1)$ depends on $m, x^{(0)}$, line search type

- very simple, but often very slow; rarely used in practice


## quadratic problem in $\mathbf{R}^{2}$

$$
f(x)=(1 / 2)\left(x_{1}^{2}+\gamma x_{2}^{2}\right)
$$

with exact line search, starting at $x^{(0)}=(\gamma, 1)$ :

$$
x_{1}^{(k)}=\gamma\left(\frac{\gamma-1}{\gamma+1}\right)^{k}, \quad x_{2}^{(k)}=\left(-\frac{\gamma-1}{\gamma+1}\right)^{k}
$$

- very slow if $\gamma \gg 1$ or $\gamma \ll 1$
- example for $\gamma=10$ :


