
Algorithm 1: Richardson's method

x_0 given
 $r_0 = b - A x_0$
 $k = 0$
repeat
 $p_k = r_k$
 $\alpha_k = \tau$
 $x_{k+1} = x_k + \alpha_k p_k$
 $r_{k+1} = r_k - \alpha_k A p_k = b - A x_{k+1}$
 $k = k + 1$
until *stopping criterion fulfilled, e.g. $\|r_k\|_{\ell^2} \leq \varepsilon \|r_0\|_{\ell^2}$*

Algorithm 2: Method of steepest descent

x_0 given
 $r_0 = b - A x_0$
 $k = 0$
repeat
 $p_k = r_k$
 $\alpha_k = \frac{(r_k, p_k)}{(A p_k, p_k)}$
 $x_{k+1} = x_k + \alpha_k p_k$
 $r_{k+1} = r_k - \alpha_k A p_k = b - A x_{k+1}$
 $k = k + 1$
until *stopping criterion fulfilled, e.g. $\|r_k\|_{\ell^2} \leq \varepsilon \|r_0\|_{\ell^2}$*

Algorithm 3: Conjugate gradient method

x_0 given
 $r_0 = b - A x_0$
 $k = 0$
repeat
 if $k = 0$ then
 $p_k = r_k$
 else
 $\beta_{k-1} = -\frac{(r_k, A p_{k-1})}{(p_{k-1}, A p_{k-1})}$
 $p_k = r_k + \beta_{k-1} p_{k-1}$
 end
 $\alpha_k = \frac{(r_k, p_k)}{(A p_k, p_k)}$
 $x_{k+1} = x_k + \alpha_k p_k$
 $r_{k+1} = r_k - \alpha_k A p_k = b - A x_{k+1}$
 $k = k + 1$
until *stopping criterion fulfilled, e.g. $\|r_k\|_{\ell^2} \leq \varepsilon \|r_0\|_{\ell^2}$*
