

## Series 4

In the last exercise, you implemented various regularization methods for the direct operator  $T$  defined by  $(Tf)' = f$  (see Series 3 for details). We will now consider one more method, CG, and then use the discrepancy principle as an a-posteriori parameter choice rule.

### 1. Conjugate Gradient Method

Implement the Conjugate Gradient method for solving  $T^*Tf = T^*g^\delta$ .

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**Algorithm 1** Conjugate Gradient Algorithm for  $T^*Tf = T^*g^\delta$

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 $f_0 = 0, r_0 = T^*g^\delta, p_0 = r_0, k = 0$   
while  $\|g^\delta - Tf_k\|_Y > tol$  do  
   $\alpha = (r_k, r_k)_X / (Tp_k, Tr_k)_Y$   
   $f_{k+1} = f_k + \alpha p_k$   
   $r_{k+1} = r_k - \alpha T^*Tp_k$   
   $\beta = (Tr_{k+1}, Tp_k)_Y / (Tp_k, Tp_k)_Y$   
   $p_{k+1} = r_{k+1} - \beta p_k$   
   $k = k + 1$   
end while
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### 2. Discrepancy Principle

For all exercises below, consider the perturbed data  $g^\delta(x) = \sin(2\pi x) + \delta \frac{\chi_{[0.4, 0.6]}(x)}{\sqrt{2}}$ .

A common choice for the parameter  $\tau$  in the discrepancy principle is 2.

- a) **Tikhonov** Implement the Newton method elucidated in Section 1.4.2 of the lecture to compute the regularization parameter  $\alpha_d$ . For  $\delta \in \{0.1, 0.05, 0.01, 0.005, 0.001\}$ , compute the Tikhonov solution using  $\alpha_d$  and plot the error vs  $\delta$ .
- b) **Landweber** Modify your implementation of the Landweber iteration to stop once the condition imposed by the discrepancy principle is met, i.e.  $\|g^\delta - T\hat{f}_k\| \leq \tau\delta$ . Plot the error vs.  $\delta$ .
- c) **Conjugate Gradients** Modify your implementation of the Conjugate Gradient method to stop once the condition imposed by the discrepancy principle is met, i.e.  $\|g^\delta - T\hat{f}_k\| \leq \tau\delta$ . Plot the error vs.  $\delta$ .