

## Problem 1

- i) Yes, the cost is strongly convex. Write  $g(x) = \frac{1}{2} \|x\|_2^2$ , and  $h_t(x) = Ax - b_t$ . Then  $g$  is  $\mu = 1$ -strongly convex, and  $h_t$  is affine for all  $t$ , so our cost function is the composition  $f = g \circ h_t$ , which is also 1-strongly convex.
- ii) We know from class that for convergence, we require the step size  $\alpha$  to be in the range  $(0, 2/L]$ . Since  $L$  is the Lipschitz constant of  $\nabla f = Ax - b_t$ , we have that since

$$\|Ax - b_t - (Ay - b_t)\| = \|Ax - Ay\| \leq \|A\| \|x - y\|,$$

that  $L \leq \|A\| = 1$ , since the largest singular value of  $A$  is 1. Thus we must pick  $\alpha \in (0, 2]$ . The bound we derived in class is in terms of

$$\rho = \max\{|1 - \alpha\mu|, |1 - \alpha L|\} = |1 - \alpha|$$

since  $\mu = L = 1$  for our problem. Then the bound at step  $t$  is

$$\|x_t - x_t^*\| \leq \rho^t \|x_0 - x_0^*\| + \sigma \sum_{i=0}^t \rho^i = \rho^t \|x_0 - x_0^*\| + \frac{1 - \rho^t}{1 - \rho}$$

with the asymptotic result

$$\limsup_{t \rightarrow \infty} \|x_t - x_t^*\| \leq \frac{1}{1 - \rho}.$$

Thus we have a tradeoff: if we pick  $\alpha$  to be near 1, then  $\rho$  is small, and our asymptotic limit is good, but the initial error  $\|x_0 - x_0^*\|$  affects our estimate for longer. If we pick  $\alpha$  to be near 0, then the initial error decays rapidly, but the asymptotic error is controlled very weakly.