October 28, 2019

0.1 Question 4 (8 marks): implement accelerated gradient for the Total-Variation denoising problem. Use the pseudo-Huber function to smooth the problem. Use the Lipschitz constant that you obtained in Q3. Do not include computation of the Lipschitz constant in this question. You can do it in Q3 and the time for computing the Lipschitz constant will not be taken into account.

```
[]: def accelerated_gd_lip(x0, epsilon, lambda_, mu, max_iterations):
       z = x0.copy().toarray()
       x_updated = x0.copy().toarray()
       f vals = []
       norm_vals = []
       t1 = time.time()
       alpha = 1.0 / L
       lamb = 1.0
       for i in range(1, max_iterations+1):
           gamma = 2.0 / i if i > 4 else 0
           y = (1.0 - gamma) * x_updated + gamma * z
           current_grad = grad_f_tv2(lamb=lambda_, mu=mu, x=y, z_n=noisy_image_vec)
           current_grad_norm = np.linalg.norm(current_grad)
           if current_grad_norm <= epsilon:</pre>
                break
           norm_vals.append(current_grad_norm)
           z = z - alpha * (gamma / lamb) * current grad
           f_vals.append(f_tv(lamb=lambda_, mu=mu, x=x_updated,__
    →z_n=noisy_image_vec))
           x_updated = y - alpha * current_grad
           lamb = (1.0 - gamma) * lamb
           f_diff = (f_vals[-1] - f_vals[-2]) if len(f_vals) > 1 else None
           grad_diff = (norm_vals[-1] - norm_vals[-2]) if len(norm_vals) > 1 else_
    \hookrightarrowNone
           print(f"Step = {i}: alpha = {alpha}, gamma = {gamma}, lambda = {lamb}, __
    →Function = {f_vals[-1]}, Function Diff. = {f_diff}, Grad. Norm = _
    →{norm_vals[-1]}, Grad. Norm. Diff. = {grad_diff}")
       t2 = time.time()
       print(f"Iterations (Total) time = {t2-t1}")
       return x_updated, np.array(f_vals)
```

0.2 Call accelerated gradient to denoise the image. Use the same λ and μ that you used in Q1.

```
epsilon=epsilon,

⊔

→max_iterations=max_iterations)
```