

Math in project1-Lipschitz Bound

1) Definition :

A differentiable function f is L -smooth or has a Lipschitz continuous gradient with constant $L > 0$ if

$$\|\nabla f(x) - \nabla f(z)\| \leq L\|x - z\|$$

for any x and z . The norm is ℓ_2 norm.

2) Lipschitz for this objective function

Our objective is:

$$f(x) = \frac{1}{2}\|k * x - y\|^2 + \frac{\lambda}{2}\|x\|^2$$

In circular convolution, we can use a circulant matrix to represent the convolution, which is

$$Ax := k * x$$

$$f(x) = \frac{1}{2}\|Ax - y\|^2 + \frac{\lambda}{2}\|x\|^2$$

We take the derivative:

$$\nabla f(x) = A^\top(Ax - y) + \lambda x$$

which is the comments in the code:

```
grad = k^T*(k*x - y) + lam*x
```

Using definition of Lipschitz to get L value

According to the definition, we need to calculate the difference:

$$\nabla f(x) - \nabla f(z) = A^\top(Ax - y) + \lambda x - (A^\top(Az - y) + \lambda z)$$

$$\nabla f(x) - \nabla f(z) = A^\top A(x - z) + \lambda(x - z) = (A^\top A + \lambda I)(x - z)$$

Take the norm:

$$|\nabla f(x) - \nabla f(z)| = |(A^\top A + \lambda I)(x - z)| \leq |A^\top A + \lambda I|_2 |x - z|$$

Therefore, we get:

$$L = |A^\top A + \lambda I|_2$$

Since the $A^\top A + \lambda I$ is semidefinite, thus the l2 norm should be the largest eigen value.

$$L = \lambda_{\max}(A^\top A + \lambda I) = \lambda_{\max}(A^\top A) + \lambda$$

Proof:

assume we have one eigenvector v for $A^\top A$

$$A^\top A v = \mu v$$

Then we have:

$$(A^\top A + \lambda I)v = A^\top A v + \lambda I v = \mu v + \lambda v = (\mu + \lambda)v$$

Thus:

$$\lambda_{\max}(A^\top A + \lambda I) = \max_i (\mu_i + \lambda) = \max_i \mu_i + \lambda = \lambda_{\max}(A^\top A) + \lambda$$

In circular convolution, we have:

$$A^\top A = (F^{-1} \text{diag}(K) F)^\top (F^{-1} \text{diag}(K) F) = F^{-1} \text{diag}(\overline{K}) \text{diag}(K) F = F^{-1} \text{diag}(|K|^2) F$$

Thus, the largest eigenvalue is:

$$\lambda_{\max}(A^\top A) = \max_{\omega} |K(\omega)|^2$$

$$L = \lambda_{\max}(A^\top A) + \lambda = \max_{\omega} |K(\omega)|^2 + \lambda$$

which is the code:

```
L = float(np.max(np.abs(K) ** 2).real + lam)
```

Use hessian matrix to understand

L is the upper bound of the largest eigenvalue of hessian matrix!

$$\left\| \frac{\nabla f(\mathbf{x}) - \nabla f(\mathbf{z})}{\mathbf{x} - \mathbf{z}} \right\| \leq \frac{\|\nabla f(\mathbf{x}) - \nabla f(\mathbf{z})\|}{\|\mathbf{x} - \mathbf{z}\|} \leq L$$

$$\|\nabla^2 f(x)\| \leq L$$

$$\nabla^2 f(x) = H(x) = \left[\frac{\partial^2 f}{\partial x_i \partial x_j} \right]_{i,j=1}^n$$