TD: Gradient Descent for convex and smooth functions

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week 4-5 - Nov. 28 (lecture). Dec 05. 2024

Convergence Analysis

We study the convergence for a fixed step size α . Prove the following result.

Theorem Assume that $f : \mathbb{R}^n \to \mathbb{R}$ is convex and L-smooth. If x^* is a critical point of f, i.e., $\nabla f(x^*) = 0$, then the the sequence $\{x^{(k)}\}$ generated by gradient descent

$$
x^{(k+1)} = x^{(k)} + \alpha \nabla f(x^{(k)}),
$$

with fixed step size $0 \le \alpha \le \frac{1}{L}$ satisfies:

$$
f(x^{(k)}) - f(x^*) \le \frac{\|x^{(0)} - x^*\|^2}{2\alpha k}.
$$

Theorem Assume that $f : \mathbb{R}^n \to \mathbb{R}$ is convex and L-smooth. If x^* is a critical point of f, i.e., $\nabla f(x^*) = 0$, then the sequence $\{x^{(k)}\}_{k=0}^{\infty}$ generated by a gradient descent

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x^{(k+1)} = x^{(k)} + \alpha \nabla f(x^{(k)}),
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with fixed step size $\alpha \leq \frac{1}{L}$ satisfies:

$$
f(x^{(k)}) - f(x^*) \le \frac{\|x^{(0)} - x^*\|^2}{2\alpha k}.
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i.e., This implies that gradient descent has a convergence rate of $O\left(\frac{1}{k}\right)$. i.e., To achieve $f(x^{(k)}) - f(x^*) \leq \epsilon$, we need $O\left(\frac{1}{\epsilon}\right)$ iterations.

Proof: Using the smoothness property, we can write:

$$
f(y) \le f(x) + \nabla f(x)^\top (y - x) + \frac{L}{2} \|y - x\|^2
$$
 for any x, y

Proving this property:

• Since f is L-smooth, then ∇f is L-Lipschitz continuous, this means there exists a constant $L > 0$ such that

$$
\nabla^2 f \preceq L I, \quad \text{or equivalently,} \quad \nabla^2 f(z) - L I \preceq 0
$$

• i.e., $\nabla^2 f(z) - LI$ is semi-definite negative, which means $\forall x, y, z$ we have:

$$
(x - y)^{\top} (\nabla^2 f(z) - LI)(x - y) \le 0
$$

which means:

$$
(x - y)^{\top} \nabla^2 f(z) (x - y) = (x - y)^{\top} \nabla^2 f(z) (x - y) - L ||x - y||^2 \le 0
$$

Rearranging this inequality, we get the bound:

$$
(x - y)^{\top} \nabla^2 f(z)(x - y) \le L \|x - y\|^2
$$

• Based on Taylor's Remainder Theorem, we have $\forall x, y, \exists z \in [x, y]$:

$$
f(y) = f(x) + \nabla f(x)^{\top} (y - x) + \frac{1}{2} (x - y)^{\top} \nabla^2 f(z) (x - y)
$$

where $\nabla f(x)$ is the gradient of f at x, $\nabla^2 f(z)$ is the Hessian matrix of f evaluated at some intermediate point $z \in [x, y]$, and the notation $z \in [x, y]$ (i.e., z lies on the line segment between x and y, i.e., $z = x + t(y - x)$ for some $t \in (0,1)$.

• Substituting the bound from the previous step into Taylor's expansion, we get:

$$
f(y) \le f(x) + \nabla f(x)^\top (y - x) + \frac{L}{2} \|y - x\|^2
$$
 for any x, y

• Plugging in $y = x^{(k+1)}$ and $x = x^{(k)}$ with $x^{(k+1)} = x^{(k)} - \alpha \nabla f(x^{(k)})$. To simplify notation, let's use $x^+ = x - \alpha \nabla f(x)$:

$$
f(x^{+}) \leq f(x) + \nabla f(x)^{\top}(x^{+} - x) + \frac{L}{2}||x^{+} - x||^{2}
$$

= $f(x) + \nabla f(x)^{\top}(x - \alpha \nabla f(x) - x) + \frac{L}{2}||x - \alpha \nabla f(x) - x||^{2}$
= $f(x) - \alpha \nabla f(x)^{\top} \nabla f(x) + \frac{L}{2} \alpha^{2} ||\nabla f(x)||^{2}$
= $f(x) - \left(1 - \frac{L\alpha}{2}\right) \alpha ||\nabla f(x)||^{2}$

• Taking $0 < \alpha \leq \frac{1}{L}$, we have $1 - \frac{L\alpha}{2} \geq \frac{1}{2}$. Therefore:

$$
f(x^{+}) \le f(x) - \frac{\alpha}{2} \|\nabla f(x)\|^{2}.
$$

• Since f is convex, $f(x) \leq f(x^*) + \nabla f(x)^\top (x - x^*)$, we have:

$$
f(x^{+}) \leq f(x) - \frac{\alpha}{2} \|\nabla f(x)\|^{2}
$$

\n
$$
\leq f(x^{*}) + \nabla f(x)^{\top}(x - x^{*}) - \frac{\alpha}{2} \|\nabla f(x)\|^{2}
$$

\n
$$
= f(x^{*}) + \frac{1}{2\alpha} \left(2\alpha \nabla f(x)^{\top}(x - x^{*}) - \alpha^{2} \|\nabla f(x)\|^{2}\right)
$$

• using the fact that $2\alpha \nabla f(x)^\top (x - x^*) - \alpha^2 \|\nabla f(x)\|^2$ is a part of a remarkable identity $||a - b||^2 = ||a||^2 - 2a^\top b + ||b||^2$ where

$$
a = x - x^*, \quad b = \alpha \nabla f(x),
$$

since $||x - x^* - \alpha \nabla f(x)||^2 = ||x - x^*||^2 - 2\alpha(x - x^*)^\top \nabla f(x) + \alpha^2 ||\nabla f(x)||^2$. Then we have

$$
2\alpha(x - x^*)^{\top}\nabla f(x) - \alpha^2 \|\nabla f(x)\|^2 = \|x - x^*\|^2 - \|x - x^* - \alpha \nabla f(x)\|^2.
$$

• The previous inequality becomes

$$
f(x^{+}) \leq f(x) - \frac{\alpha}{2} \|\nabla f(x)\|^{2}
$$

\n
$$
\leq f(x^{*}) + \frac{1}{2\alpha} (\|x - x^{*}\|^{2} - \|x - x^{*} - \alpha \nabla f(x)\|^{2})
$$

\n
$$
= f(x^{*}) + \frac{1}{2\alpha} (\|x - x^{*}\|^{2} - \|x^{+} - x^{*}\|^{2})
$$

and we finally get

$$
f(x^{+}) - f(x^{*}) \le \frac{1}{2\alpha} (||x - x^{*}||^{2} - ||x^{+} - x^{*}||^{2})
$$

• This inequality holds for x^+ on every iteration of gradient descent. Summing over iterations, we have:

$$
\sum_{i=1}^{k} \left(f(x^{(i)}) - f(x^*) \right) \leq \sum_{i=1}^{k} \frac{1}{2\alpha} \left(\|x^{(i-1)} - x^*\|_2^2 - \|x^{(i)} - x^*\|_2^2 \right)
$$

telescoping series $\frac{1}{2\alpha} \left(\|x^{(0)} - x^*\|_2^2 - \|x^{(k)} - x^*\|_2^2 \right)$
 $\leq \frac{1}{2\alpha} \left(\|x^{(0)} - x^*\|_2^2 \right)$

So we obtain:

$$
\sum_{i=1}^{k} \left(f(x^{(i)}) - f(x^*) \right) \le \frac{1}{2\alpha} \|x^{(0)} - x^*\|_2^2
$$

• Since $f(x^{(k)})$ is nonincreasing,

$$
kf(x^{(k)}) \leq \sum_{i=1}^{k} f(x^{(i)})
$$

which implies

$$
k(f(x^{(k)}) - f(x^*)) \le \sum_{i=1}^k (f(x^{(i)}) - f(x^*)),
$$

equivalently,

$$
f(x^{(k)}) - f(x^*) \le \frac{1}{k} \sum_{i=1}^k (f(x^{(i)}) - f(x^*)).
$$

Thus:

$$
f(x^{(k)}) - f(x^*) \le \frac{1}{k} \sum_{i=1}^k \left(f(x^{(i)}) - f(x^*) \right) \le \frac{\|x^{(0)} - x^*\|^2}{2\alpha k}
$$

We then finally have:

$$
f(x^{(k)}) - f(x^*) \le \frac{\|x^{(0)} - x^*\|^2}{2\alpha k}.
$$

appendix

Telescoping Series: To understand why

$$
\sum_{i=1}^{k} \frac{1}{2\alpha} \left(\|x^{(i-1)} - x^*\|_2^2 - \|x^{(i)} - x^*\|_2^2 \right) = \frac{1}{2\alpha} \left(\|x^{(0)} - x^*\|_2^2 - \|x^{(k)} - x^*\|_2^2 \right)
$$

let's expand the summation to observe the telescoping effect:

$$
\sum_{i=1}^{k} \frac{1}{2\alpha} \left(\|x^{(i-1)} - x^*\|_2^2 - \|x^{(i)} - x^*\|_2^2 \right)
$$

Expanding this explicitly, we have:

$$
\frac{1}{2\alpha} \left(\|x^{(0)} - x^*\|^2_2 - \|x^{(1)} - x^*\|^2_2 \right) + \frac{1}{2\alpha} \left(\|x^{(1)} - x^*\|^2_2 - \|x^{(2)} - x^*\|^2_2 \right) + \cdots
$$

$$
+ \frac{1}{2\alpha} \left(\|x^{(k-1)} - x^*\|^2_2 - \|x^{(k)} - x^*\|^2_2 \right)
$$

Notice that most intermediate terms cancel:

- The term $||x^{(1)} x^*||_2^2$ appears as a positive value in the first part and cancels with the negative value in the next part.
- Similarly, the term $||x^{(2)} x^*||_2^2$ cancels out, and this pattern continues.

Thus, the only terms that do not cancel are the **first** term $||x^{(0)} - x^*||_2^2$ and the **ast** negative term $-\Vert x^{(k)} - x^* \Vert_2^2$, which results in:

$$
\frac{1}{2\alpha}\left(\|x^{(0)}-x^*\|_2^2-\|x^{(k)}-x^*\|_2^2\right)
$$