

Sofar: Convergence properties on
CVX problems

This lecture: Some guarantees
for some nonconvex problems

Q: What can we say about general
non-CVX problems?

A: Not much because $f(w)$
can have:
• exp. num of local mins
• " " " saddle points

But, we can show convergence
to critical point, or add structure
and prove bounds for some non-CVX losses

Q: Why can't we show global convergence?

A: In general $\min_w f(w)$ can encode NP-Hard problems.

What about local convergence?

SGD on smooth non-convex fun:

Reminder:

f is β -smooth if

$$\|\nabla f(x) - \nabla f(y)\| \leq \beta \|x-y\|, \forall x, y \in \mathbb{R}^d$$

This implies:

$$|f(x) - f(y) - \langle \nabla f(y), x-y \rangle| \leq \frac{\beta}{2} \|x-y\|^2, \forall x, y \in \mathbb{R}^d$$

Let $x = \omega_{k+1}, y = \omega_k$

$$\omega_{k+1} = \omega_k - \gamma \nabla f_{S_k}(\omega_k)$$

Then smoothness implies

$$f(\omega_{k+1}) - f(\omega_k) - \langle \nabla f(\omega_k), \omega_{k+1} - \omega_k \rangle \leq \beta \|\gamma \nabla f_{S_k}(\omega_k)\|^2$$

$$\Rightarrow f_{k+1} \leq f_k - \gamma \langle \nabla f(\omega_k) \nabla f_{S_k}(\omega_k) \rangle + \beta \gamma^2 \|\nabla f_{S_k}(\omega_k)\|^2$$

$$\Rightarrow |E f_{k+1}| \leq |E f_k - \gamma |E \|\nabla f(\omega_k)\|^2 + \beta \gamma^2 \underbrace{\|\nabla f_{S_k}(\omega_k)\|^2}_{\leq M^2}$$

$$\leq |E f_k - \gamma |E \|\nabla f(\omega_k)\|^2 + \beta \gamma^2 M^2$$

$$\Rightarrow \gamma |E \|\nabla f(\omega_k)\|^2 \leq |E (f_k - f_{k+1})| + \beta \gamma^2 M^2$$

$$\Rightarrow E \|\nabla f(\omega_k)\|^2 \leq \frac{|E (f_k - f_{k+1})|}{\gamma} + \beta \gamma M^2$$

Let's now sum for T iterations

$$\mathbb{E} \|\nabla f(\omega_1)\|^2 \leq \frac{\mathbb{E} f_1 - f^*}{\gamma} + \beta \gamma M^2$$

$$\mathbb{E} \|\nabla f(\omega_2)\|^2 \leq \frac{\mathbb{E} f_2 - f^*}{\gamma} + \beta \gamma M^2$$

\vdots

$$\sum_{k=1}^T \mathbb{E} \|\nabla f(\omega_k)\|^2 \leq \frac{\mathbb{E} f_t - f^*}{\gamma} + T \cdot \beta \gamma M^2$$

$$\Rightarrow T \cdot \min_{t=1:T} \mathbb{E} \|\nabla f(\omega_t)\|^2 \leq \frac{\mathbb{E} f_t - f^*}{\gamma} + T \beta \gamma M^2$$

$$\Rightarrow \min_{t=1:T} \mathbb{E} \|\nabla f(\omega_t)\|^2 \leq \frac{\mathbb{E} f_t - f^*}{T \gamma} + \beta \gamma M^2$$

Let's assume $f_t - f^* \leq D$

then set $\gamma = \sqrt{\frac{D}{\beta M^2 T}}$ to get:

$$\min_t \mathbb{E} \|\nabla f(\omega_t)\|^2 \leq 2 \sqrt{\frac{D \beta M^2}{T}}$$

The above implies that no matter what $f(w)$ looks like, if it is smooth it can reach an "approximate" critical point in $O(\frac{1}{\epsilon})$ iterations.

This rate seems too slow to explain practical perf. of SGD.

What if we add a bit more structure
Polyak-Łojasiewicz (PL) - condition:

A function is μ -PL if

$$\|\nabla f(x)\|^2 \geq \mu(f(x) - f^*)$$

in combination with β -smooth

$$\Rightarrow \beta^2 \|x - \Pi_{x^*}(x)\|^2 \geq \mu(f(x) - f^*)$$

That is, the furter we move from the OPT the the more the suboptimality increases.

Remark: Not all PL funs are CVX.
If a CVX fun is PL, then it's strongly-CVX.

Remark: All local minima are global.
(This happens in some NN cases)

SUD on PL funs:

From previous analysis:

$$\begin{aligned} \mathbb{E} f_{k+1} &\leq \mathbb{E} f_k - \gamma \mathbb{E} \| \nabla f(\omega_k) \|^2 + \beta \gamma^2 M^2 \\ &\leq \mathbb{E} f_k - \gamma \gamma (\mathbb{E} f_k - f^*) + \beta \gamma^2 M^2 \end{aligned}$$

Subtract f^* from both sides

$$\begin{aligned} \Rightarrow \mathbb{E} f_{k+1} - f^* &\leq (1 - \gamma \gamma) \mathbb{E} (f_k - f^*) + \beta \gamma^2 M^2 \\ &\leq (1 - \gamma \gamma)^{k+1} (f_0 - f^*) + \sum_{i=0}^k (1 - \gamma \gamma)^i \cdot \beta \gamma^2 M^2 \end{aligned}$$

$$\leq (1 - \gamma)^{k+1} D + \frac{\beta \gamma^2 \mu^2}{\gamma}$$

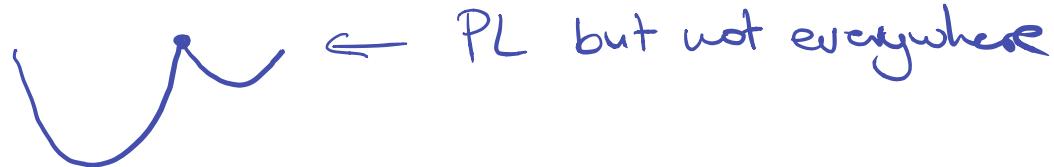
$$= \underbrace{(1 - \mu y)^{K+1} D}_{\varepsilon/2} + \frac{\beta y M^2}{\mu} \underbrace{\varepsilon}_{\varepsilon/2}$$

$$\gamma = \frac{\varepsilon \cdot h}{\beta \mu^2} \quad T = O\left(\frac{M^2 \cdot \beta}{\mu^2} \frac{1}{\varepsilon} \log \frac{D}{\varepsilon}\right)$$

Much faster conv. rate.

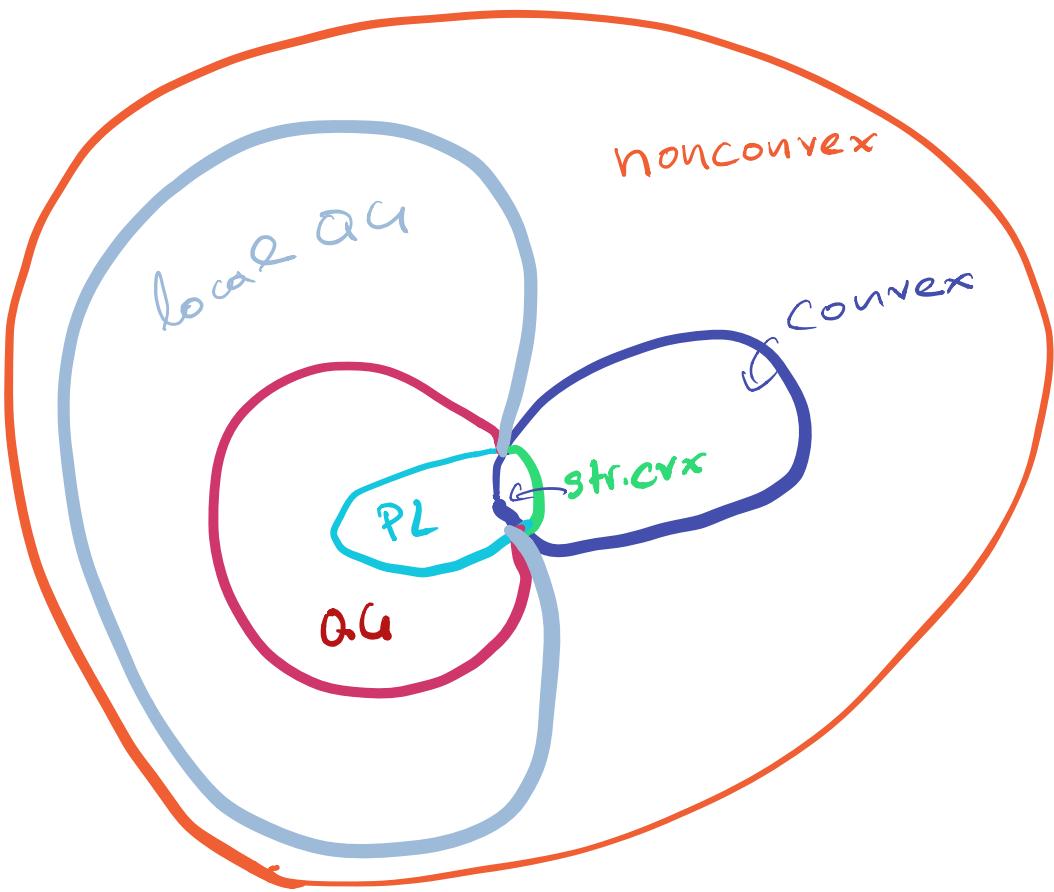
Remark: PL is a useful property but does not account for saddle points

e.g.



A more general function class: Quadratic Growth

$$f(\omega) - f^* \geq \hbar/2 \| \omega - \pi_{N^k}(\omega) \|^2$$



Unfortunately no conv. guarantees
known for QG.

Also unclear how relevant it is in practice

Several open Qs:

- What are interesting non-convex fns where we can show convergence?
- Do nns have as many local min as global min?
- What is the suboptimality of local min on real problems.

Next time:

- How to really choose stepsize?
- What happens for non-diff. function?
- What to do when we have constraints?