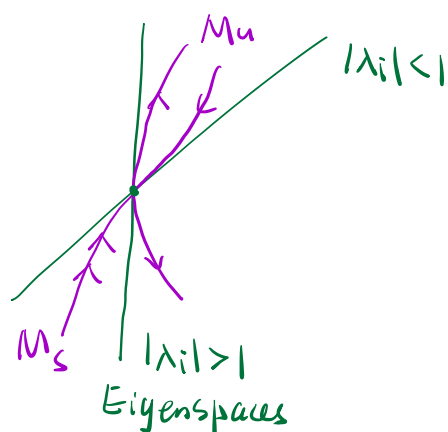


Guest Lecture by Ziang Chen

Reference: Lee et al. (2019) First order methods almost always avoid strict saddle points



strict saddle points
 $\nabla f(x^*) = 0$ $\lambda_{\min}(\nabla^2 f(x^*)) < 0$

Gradient Descent $x_{t+1} = x_t - \alpha \nabla f(x_t) = g(x_t)$
 $g = I - \alpha \nabla f$

Linearization: $A = I - \alpha H$
 $H = \nabla^2 f(x^*)$
 $\lambda_i(A) = 1 - \alpha \lambda_i(H)$

If $\lambda_i(H) < 0$, then $\lambda_i(A) > 1$

\Rightarrow the unstable manifold M_u is non-trivial
 $\dim(M_u) \geq 1$

Assume $\lambda_i(H) \neq 0, \forall i \Rightarrow |\lambda_i(A)| \neq 1, \forall i$
 \Rightarrow no center manifold

$\Rightarrow \dim(M_s) = d - \dim(M_u) \leq d-1 < d$

$\Rightarrow \text{Measure}(M_s \cap B) = 0$

If $x_t \rightarrow x^*$, ^{saddle point} then $\exists t_0$ s.t.

$$x_{t_0} \in M_s \cap B \quad \leftarrow \text{Theorem from last time}$$

$$x_{t_0-1} = g^{-1}(x_{t_0}) \in g^{-1}(M_s \cap B)$$

$$\Rightarrow x_0 \in g^{-t_0}(M_s \cap B)$$

$$\Rightarrow x_0 \in \bigcup_{t=0}^{\infty} g^{-t}(M_s \cap B)$$

$$\Rightarrow \text{measure} \left(\bigcup_{t=0}^{\infty} g^{-t}(M_s \cap B) \right)$$

$$\leq \sum_{t=0}^{\infty} \text{measure} \left(g^{-t}(M_s \cap B) \right)$$

Lemma: If $\lambda_i(Dg(x)) \neq 0$, $\forall x \in \mathbb{R}^d$
 then $\text{meas}(B) = 0 \Rightarrow \text{meas}(g^{-1}(B)) = 0$
 $= 0$

Theorem If x^* is a strict saddle,
 then $\{x_0 \mid x_t \rightarrow x^*\}$ has measure 0.

Reference: L. Arnold (1998) Random dynamical systems

Prob space $(\Omega, \mathcal{F}, \mathbb{P})$

$$\text{Ex: } \Omega = \{(\omega_0, \omega_1, \omega_2, \dots) \mid \omega_j \in \{1, \dots, d\}\}$$

$$\text{Prob}(\omega_j = k) = \frac{1}{d}, \quad \forall k \in \{1, \dots, d\}$$

Def (Metric dynamical system)

$$\theta^t = \theta \circ \theta \circ \dots \circ \theta$$

$$\theta: \Omega \rightarrow \Omega$$

$$\underline{IP(\theta^{-1}(B)) = IP(B), \quad \forall B \in \mathcal{F}}$$

probability preserving

θ is shifting operator

$$\theta(\omega) = \theta(\hat{\omega}_0, \hat{\omega}_1, \dots) = (\hat{\omega}_1, \hat{\omega}_2, \dots)$$

$$\theta(\hat{\omega}_1, \hat{\omega}_2, \dots) = (\hat{\omega}_2, \hat{\omega}_3, \dots)$$

Def (Random Dynamical System)

$$\mathcal{Q}: \mathbb{T} \times \Omega \times X \rightarrow X$$

$$(t, \omega, x) \mapsto \mathcal{Q}(t, \omega, x)$$

Cocycle property:

$$\textcircled{1} \mathcal{Q}(0, \omega, x) = x$$

$$\textcircled{2} \mathcal{Q}(t+s, \omega, x) = \mathcal{Q}(t, \theta^s \omega, \mathcal{Q}(s, \omega, x))$$

$$\forall t, s, \omega, x$$

Example: $\omega = (\hat{\omega}_0, \hat{\omega}_1, \hat{\omega}_2, \dots)$

First iteration, $g(\omega, x) = x - \eta \langle \nabla f(x), e_{i_0} \rangle e_{i_0}$, $i_0 = \pi(\omega)$

$$e_s = \begin{pmatrix} 0 \\ \vdots \\ 1 \\ \vdots \\ 0 \end{pmatrix}$$

π is an operator choosing first entry of ω

Second iteration $g(\theta\omega, x) = x - \eta \langle \nabla f(x), e_{\pi(\theta\omega)} \rangle e_{\pi(\theta\omega)}$

$$\pi(\theta\omega) = \hat{\omega}_1$$

$$x_0 \xrightarrow{g(w, \cdot)} x_1 \xrightarrow{g(\theta w, \cdot)} x_2 \longrightarrow \dots \longrightarrow x_{s-1} \xrightarrow{g(\theta^{s-1} w, \cdot)} x_s$$

$$\xrightarrow{g(\theta^{s+1} w, \cdot)} x_{s+1} \longrightarrow \dots \longrightarrow x_{t+s-1} \xrightarrow{g(\theta^{t+s-1} w, \cdot)} x_{t+s}$$