

Machine Learning Notes - 2



Lipschitz Continuity

<https://math.stackexchange.com/questions/2374289/understanding-lipschitz-continuity/2374305#2374305>

Think about the mean value theorem and Lipschitz continuity.
Mean value theorem says if f is continuous at $[a,b]$ and differentiable at (a,b) , then

$$\exists c \in (a, b) \text{ such that } \frac{f(b) - f(a)}{b - a} = f'(c).$$

Lipschitz says that

$$\exists K > 0, \forall a, b \in D_f, \text{ such that } \frac{|f(b) - f(a)|}{|b - a|} \leq K.$$

Then if the derivative of f as a function is bounded, then f will be **Lipschitz**.

Specifically, if $K == 1$, then the function is said to be **1-Lipschitz**.

Consider the case

$$f(x) = \sqrt{x} \text{ for } x \in [0, 1]$$

the f is not Lipschitz, since

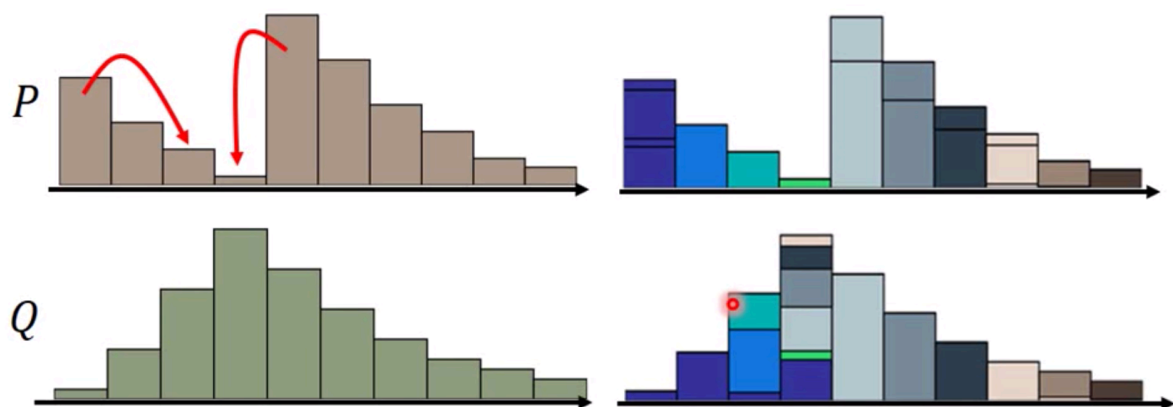
$$\sup_{x \in [0, 1]} f'(x) = \lim_{x \rightarrow 0} f'(x) = +\infty.$$

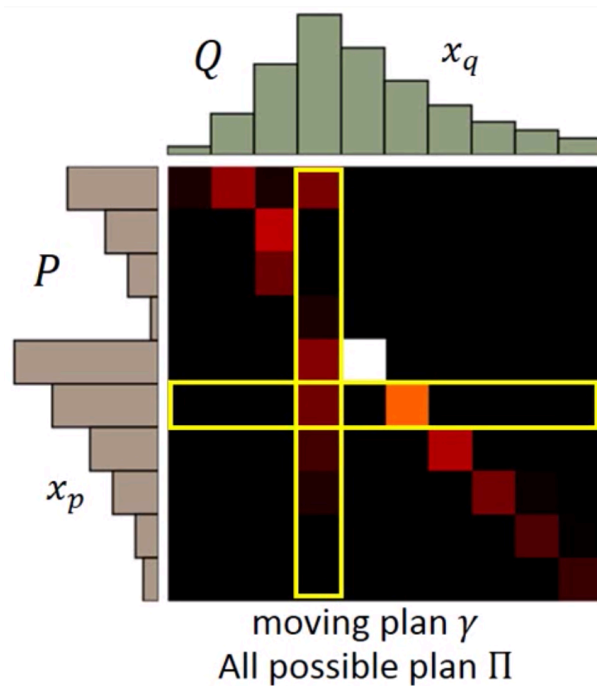
Earth Mover's Distance

<https://www.youtube.com/watch?v=KSN4QYgAtao&feature=youtu.be>

Consider one distribution **P** as a **pile of earth**, and another distribution **Q** as **the target**,

Earth Mover's Distance is the average distance the earth mover has to move the earth.





A “moving plan” is a matrix
The value of the element is the amount of earth from one position to another.

Average distance of a plan γ :

$$B(\gamma) = \sum_{x_p, x_q} \gamma(x_p, x_q) \|x_p - x_q\|$$

Earth Mover’s Distance:

$$W(\dot{P}, Q) = \min_{\gamma \in \Pi} B(\gamma)$$

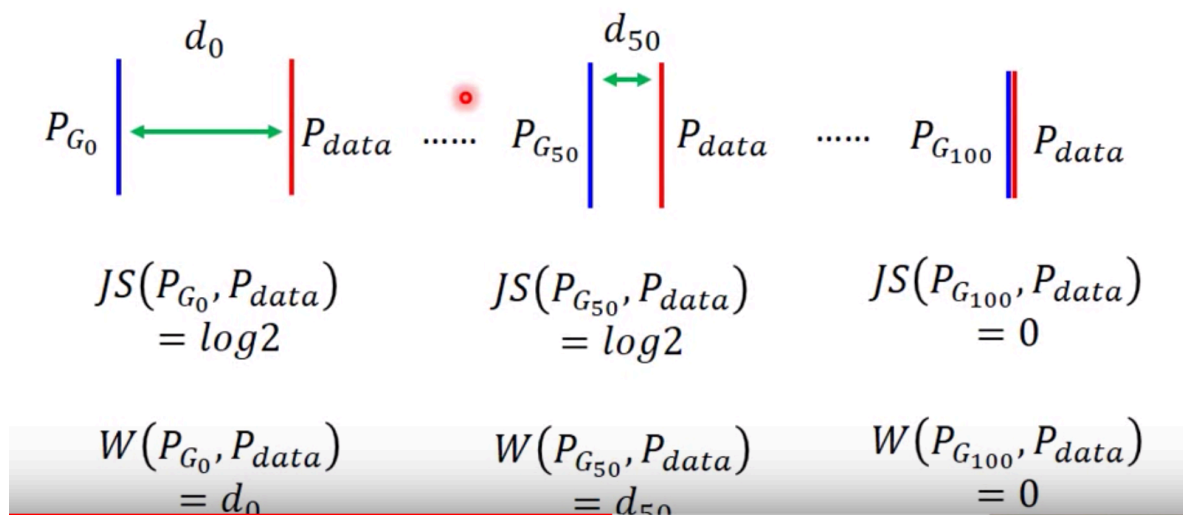
The best plan

Why Earth Mover’s Distance ?

the comparison between **JS-Divergence** and **Earth Mover’s Distance**

P_G denotes the distribution of Generator and P_{DATA} denotes the distribution of real data.

As shown below, **JSD** is less effectiveness to represent the “distance” between P_G and P_{DATA} while **Earth Mover’s Distance** does.



Back to the GAN framework

$$D_f(P_{data} || P_G) \rightarrow W(P_{data}, P_G)$$

$$= \max_D \{E_{x \sim P_{data}}[D(x)] - E_{x \sim P_G}[f^*(D(x))]\}$$

$$W(P_{data}, P_G)$$

$$= \max_{D \in 1-Lipschitz} \{E_{x \sim P_{data}}[D(x)] - E_{x \sim P_G}[D(x)]\}$$

