

Gradient descent.

1) $\Theta_0 = \Theta_{\text{init}} \sim N(\dots)$. $(x, y) \in D$ $|D| = N$

2) $\Theta_n = \Theta_{n-1} - \alpha \frac{1}{N} \sum_{i=0}^N \nabla_{\Theta} L(f(x_i, \Theta_{n-1}), y_i)$

\nwarrow
 D

1) $\Theta_0 = \Theta_{\text{init}} \sim N(\dots)$. **SGD.**

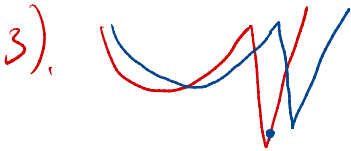
2) d is **mini batch** random subset of D of fixed size **k** .

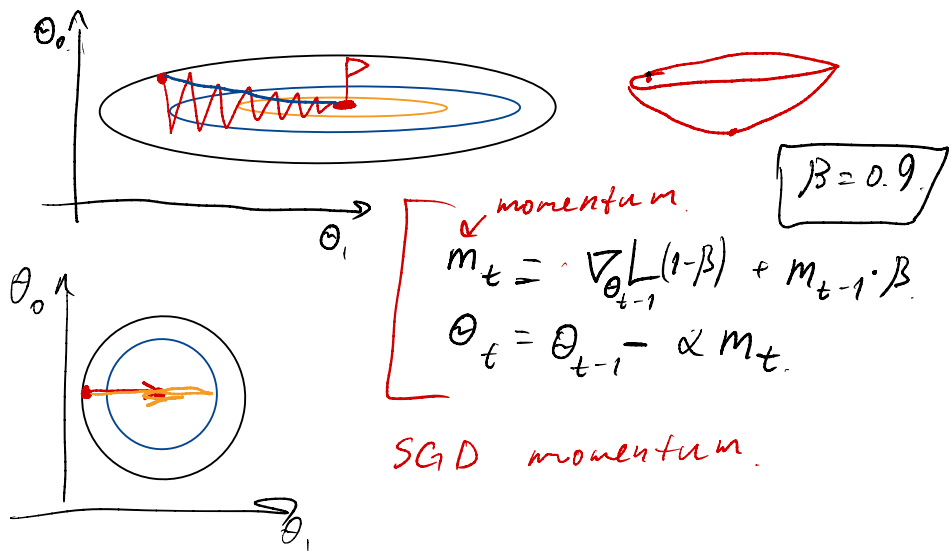
3) $\Theta_n = \Theta_{n-1} - \alpha \frac{1}{k} \sum_{(x,y) \in d} \nabla_{\Theta} L(f(x, \Theta_{n-1}), y)$

\nwarrow

1) On k subproblem "yprobems uny na".

2) $k = N$ $\alpha = \alpha_0$ $k = N/5$ $\alpha = \alpha_0/5$





$$m_t = \nabla_{\theta} L(\theta_{t-1}) \cdot (1-\beta) + m_{t-1} \cdot \beta$$

$$m_t = \nabla_{\theta} L(\theta_{t-1} + \beta m_{t-1}) \cdot (1-\beta) + m_{t-1} \cdot \beta$$

look ahead gradient.

Nesterov momentum.

$$V_t = \beta V_{t-1} + (1-\beta) (\nabla_{\theta} L(\theta_{t-1}))^2$$

$$\theta_t = \theta_{t-1} - \alpha \frac{\nabla_{\theta} L(\theta_{t-1})}{\sqrt{V_t + \epsilon}} \quad \left[\begin{array}{l} \text{AdaGrad} \\ \text{RMS Prop} \end{array} \right]$$

Ada Grad + Momentum
Ada M.

AMSgrad.

$$m_t = \nabla_{\theta} L(\theta_{t-1}) \cdot (1 - \beta_1) + m_{t-1} \cdot \beta_1$$

$$V_t = \beta_2 V_{t-1} + (1 - \beta_2) (\nabla_{\theta} L(\theta_{t-1}))^2$$

$$\theta_t = \theta_{t-1} - \alpha \frac{m_t}{\sqrt{V_t} + \epsilon}$$

$$\hat{V}_t = \beta_2 \hat{V}_{t-1} + (1 - \beta_2) (\nabla_{\theta} L(\theta_{t-1}))^2$$

$$V_t = \max(\hat{V}_t, V_{t-1})$$

Nadam

Ada Belief.

$$V_t = V_{t-1} \cdot \beta_2 + (1 - \beta_2) (m_t - \nabla_{\theta} L(\theta_{t-1}))^2$$

n - number of ops.

$\sim 2n$ $O(n)$

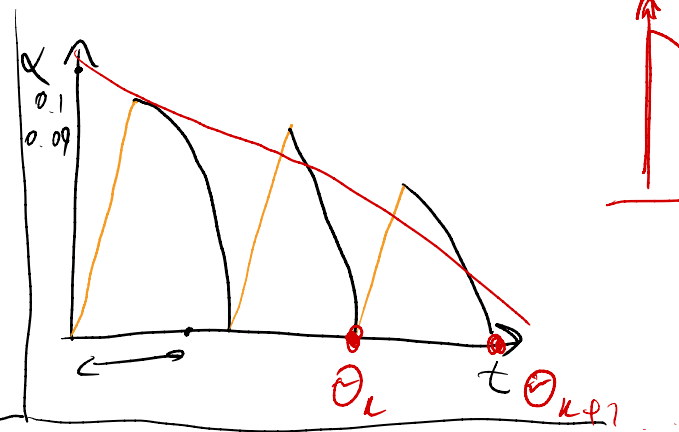
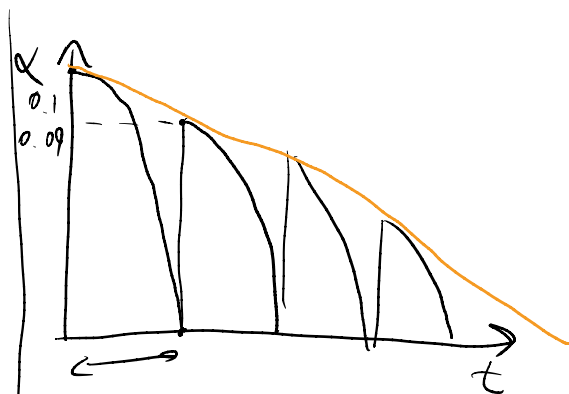
backprop.

P - num of params $|\theta|$

$\lfloor 2n \rfloor + 2p$ - momentum

$\lfloor 2n \rfloor + 3p$ - adam

$2n + 2p$ - Nesterov.



$$\Theta = \frac{1}{n} \sum \Theta_i$$