

Coursera Deep Learning Course 2-Week 2: Optimization Methods

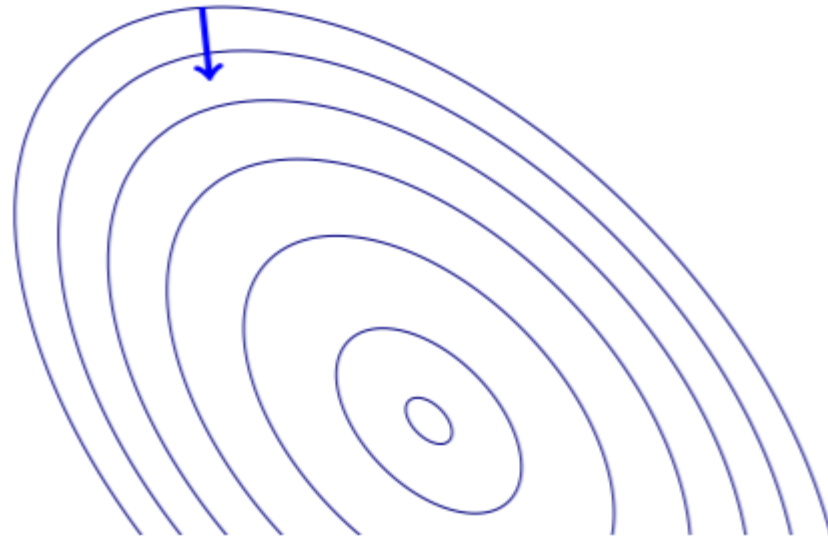
Kang Zhao

2018.09.06

Course Contents

- Mini-batch gradient descent
- Gradient Descent with momentum
- RMSprop
- Adam
- Learning Rate Decay

Optimization Overview

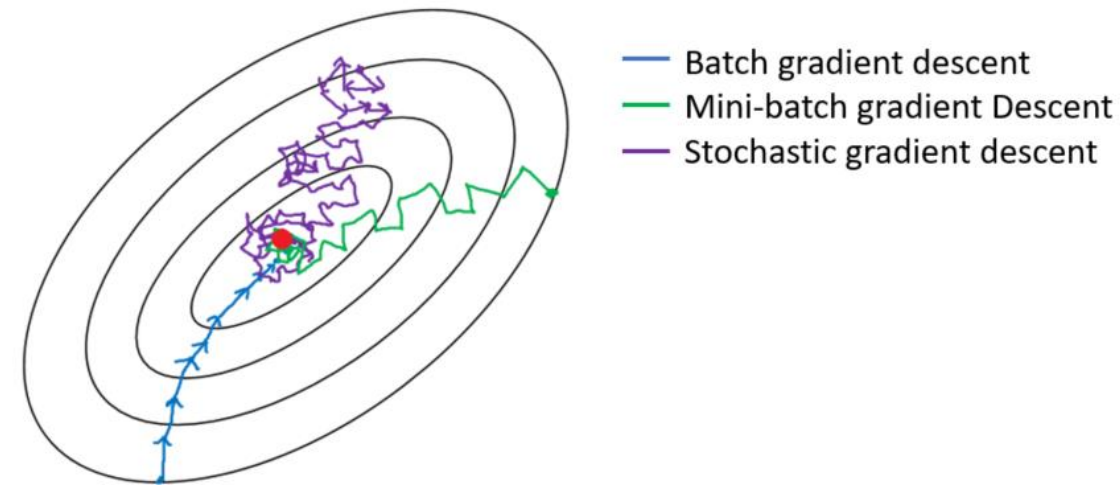


Gradient Descent

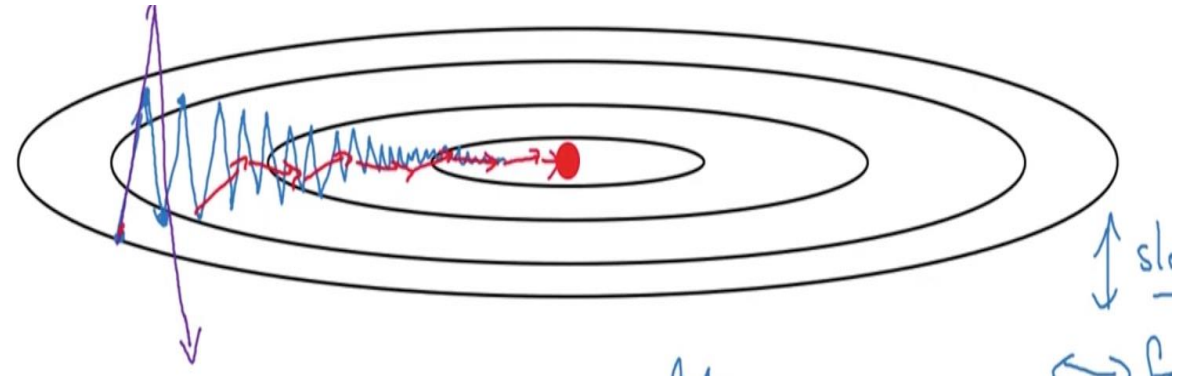
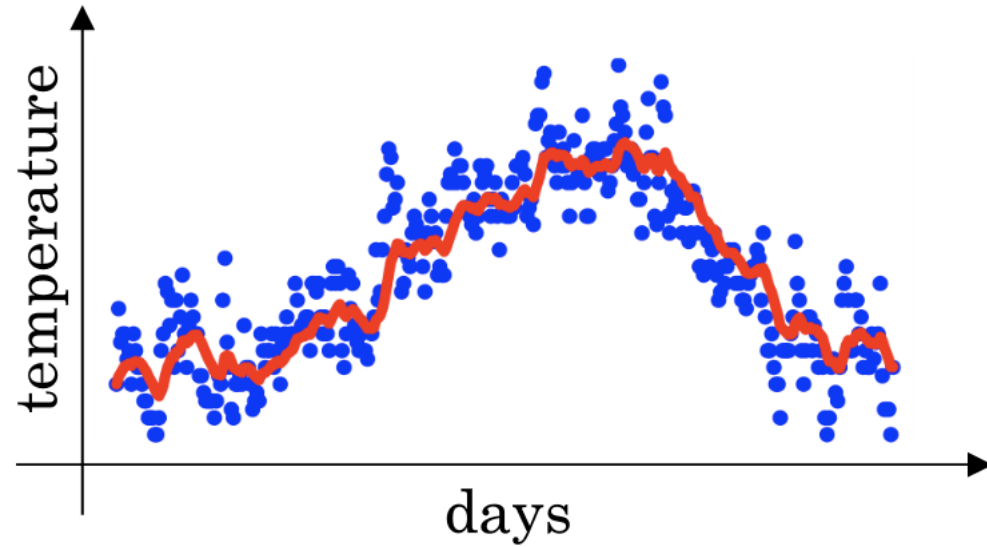
Mini-Batch & SGD

- Stochastic Gradient Descent
 - Train only one sample at one epoch
- Batch Gradient Descent
 - Train all the samples at one epoch
- Mini Batch Gradient Descent
 - Train k samples together at one epoch

Mini-batch size: K



Exponentially Weighted (Moving) Average



Momentum & RMSprop & Adam

$$\left. \begin{aligned} Vd_w &= \beta_1 * Vd_w + (1 - \beta_1) * d_w \\ Vd_b &= \beta_1 * Vd_b + (1 - \beta_1) * d_b \end{aligned} \right\} \text{“momentum”-like update}$$

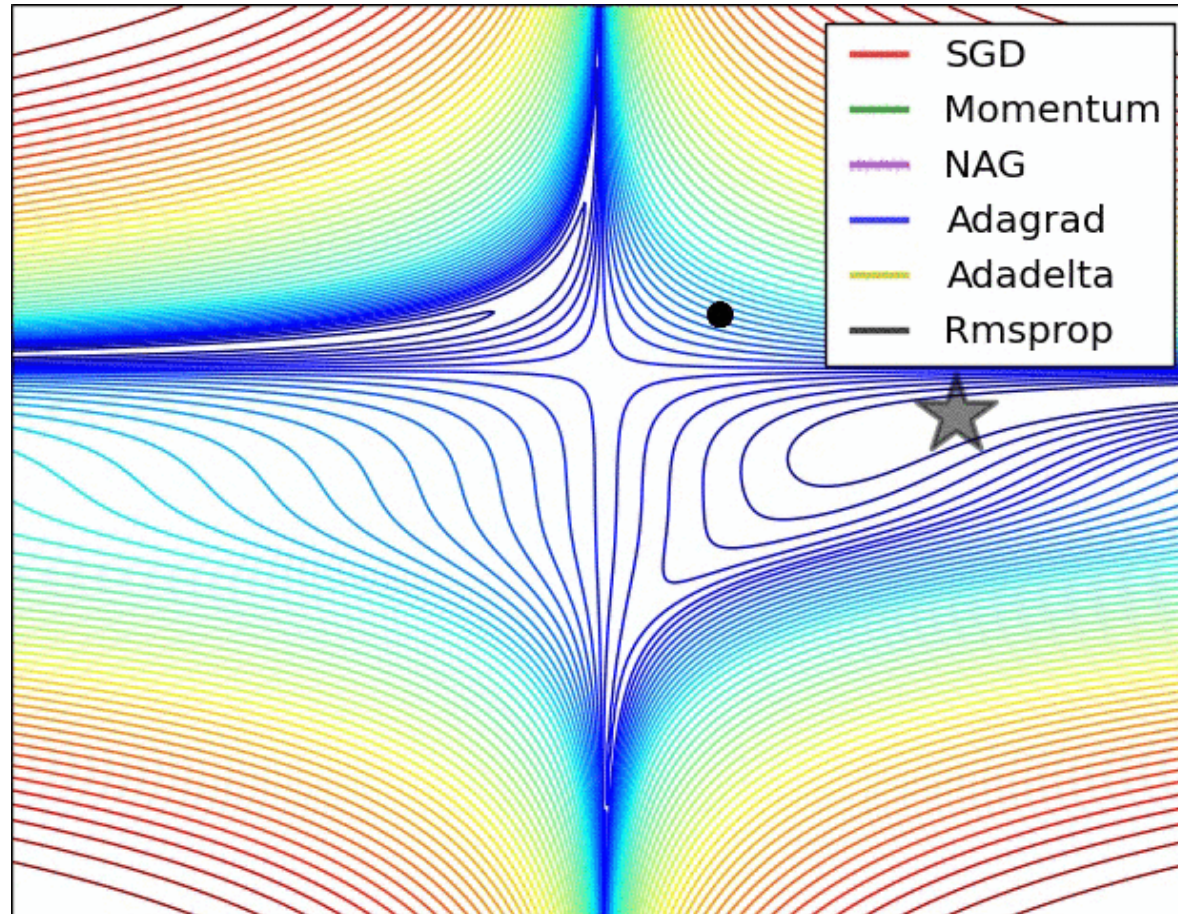
$$\left. \begin{aligned} Sd_w &= \beta_2 * Sd_w + (1 - \beta_2) * d_w^2 \\ Sd_b &= \beta_2 * Sd_b + (1 - \beta_2) * d_b^2 \end{aligned} \right\} \text{“RMSprop”}$$

$$\begin{aligned} V_{d_w}^{corrected} &= Vd_w / (1 - \beta_1^t) , & V_{d_b}^{corrected} &= Vd_b / (1 - \beta_1^t) \\ S_{d_w}^{corrected} &= Sd_w / (1 - \beta_2^t) , & S_{d_b}^{corrected} &= Sd_b / (1 - \beta_2^t) \end{aligned}$$

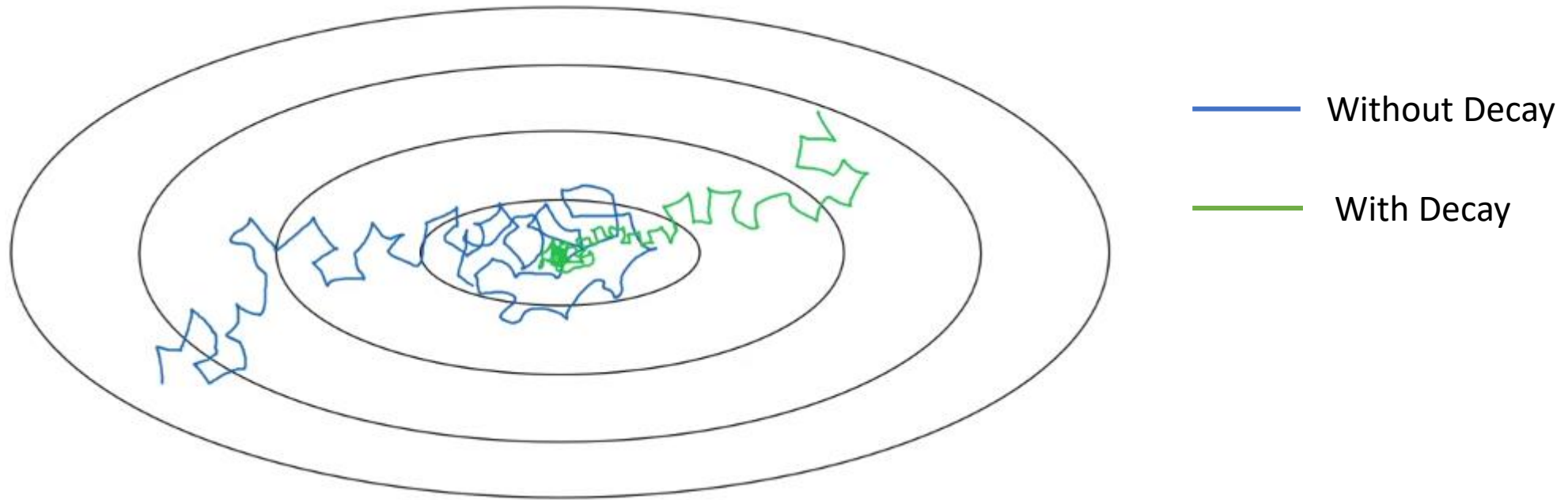
$$\begin{aligned} w &:= w - \alpha * \frac{V_{d_w}^{corrected}}{\sqrt{S_{d_w}^{corrected} + \epsilon}} \\ b &:= b - \alpha * \frac{V_{d_b}^{corrected}}{\sqrt{S_{d_b}^{corrected} + \epsilon}} \end{aligned}$$

Exponential Weight: $\beta_1, \beta_2, \epsilon$

Momentum & RMSprop & Adam



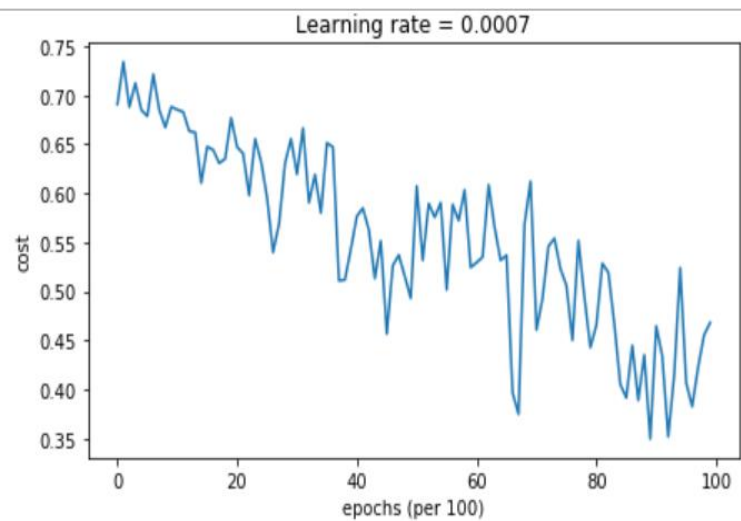
Learning Rate Decay



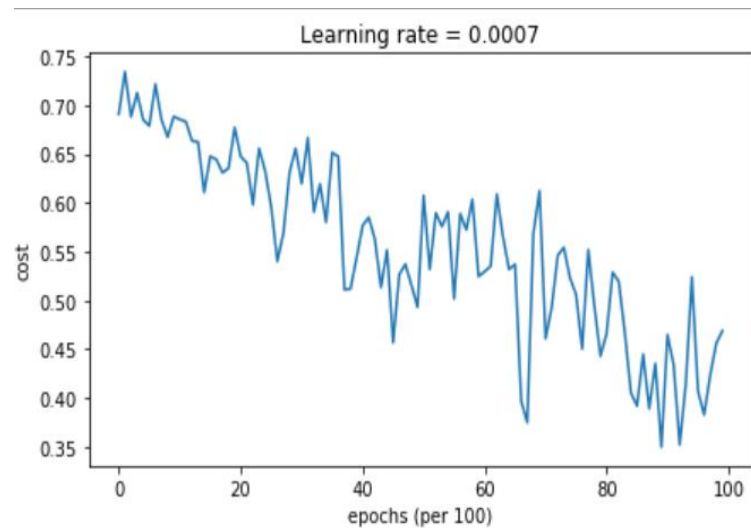
Learning Rate Decay: $\alpha = f(\alpha, epoch)$

Assignments

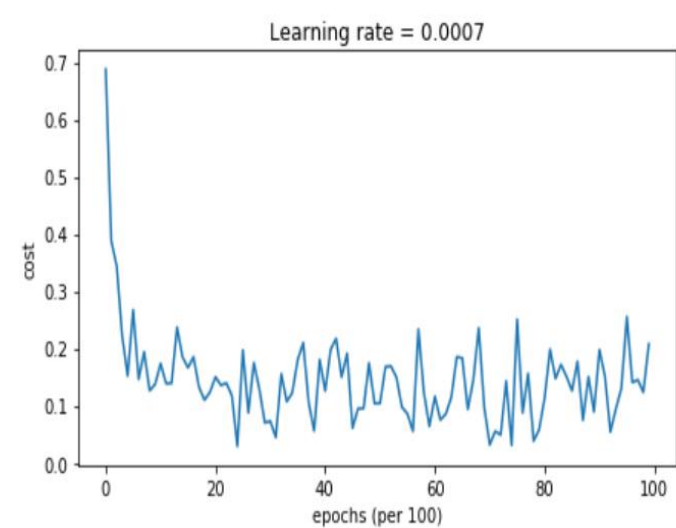
- Gradient Descent
- Mini-Batch Gradient Descent
- Momentum
- Adam
- Model with different optimization algorithms



Accuracy: 0.796666666667



ccuracy: 0.796666666667



Accuracy: 0.94

