



什么是梯度

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Clarification

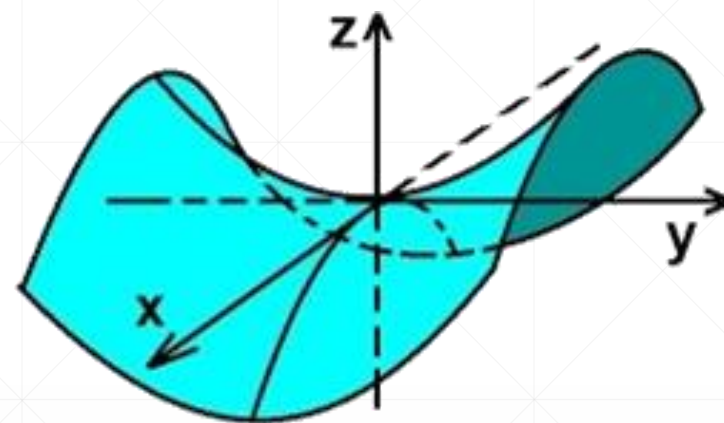
- 导数, derivate
- 偏微分, partial derivate
- 梯度, gradient

$$\nabla f = \left(\frac{\partial f}{\partial x_1}; \frac{\partial f}{\partial x_2}; \dots; \frac{\partial f}{\partial x_n} \right)$$

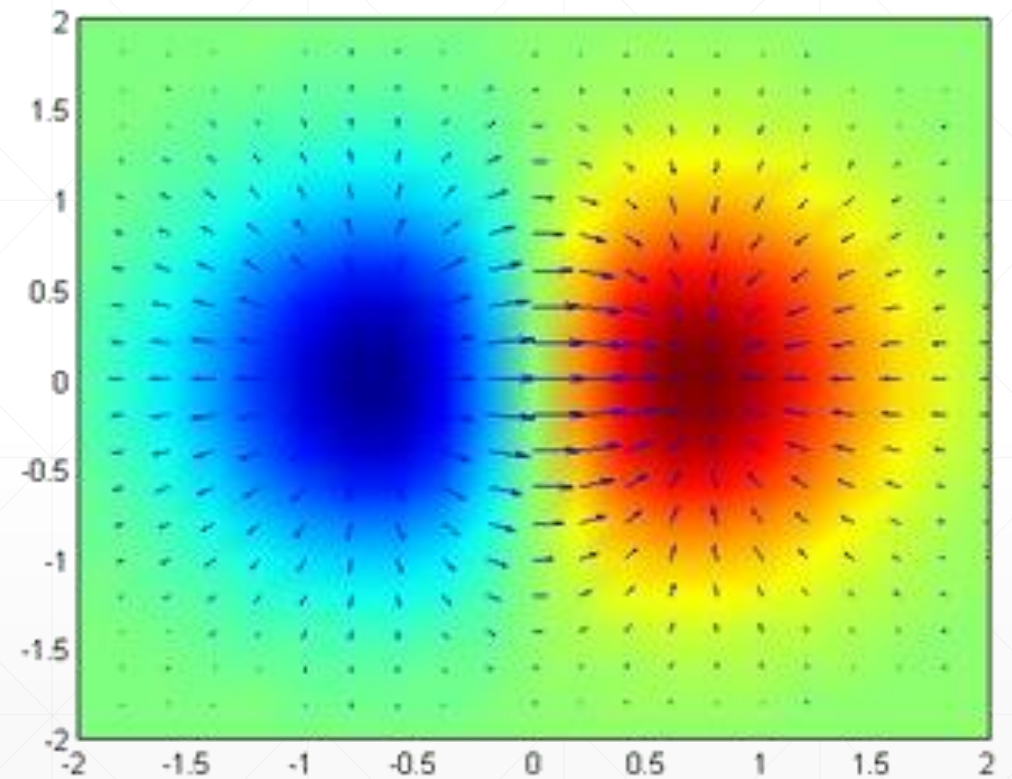
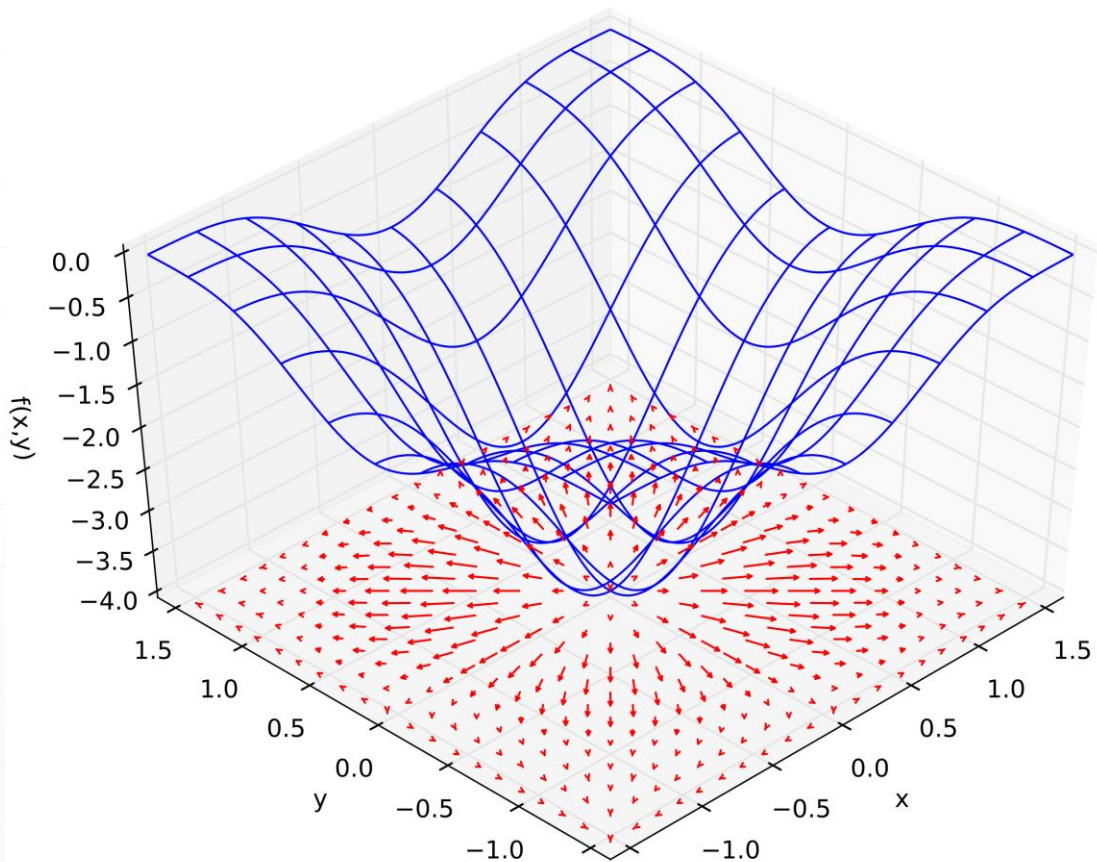
$$z = y^2 - x^2$$

$$\frac{\partial z}{\partial x} = -2x$$

$$\frac{\partial z}{\partial y} = 2y$$



What does grad mean?



How to search for minima?

$$\theta_{t+1} = \theta_t - \alpha_t \nabla f(\theta_t).$$

Function:

$$J(\theta_1, \theta_2) = \theta_1^2 + \theta_2^2$$

Objective:

$$\min_{\theta_1, \theta_2} J(\theta_1, \theta_2)$$

Update rules:

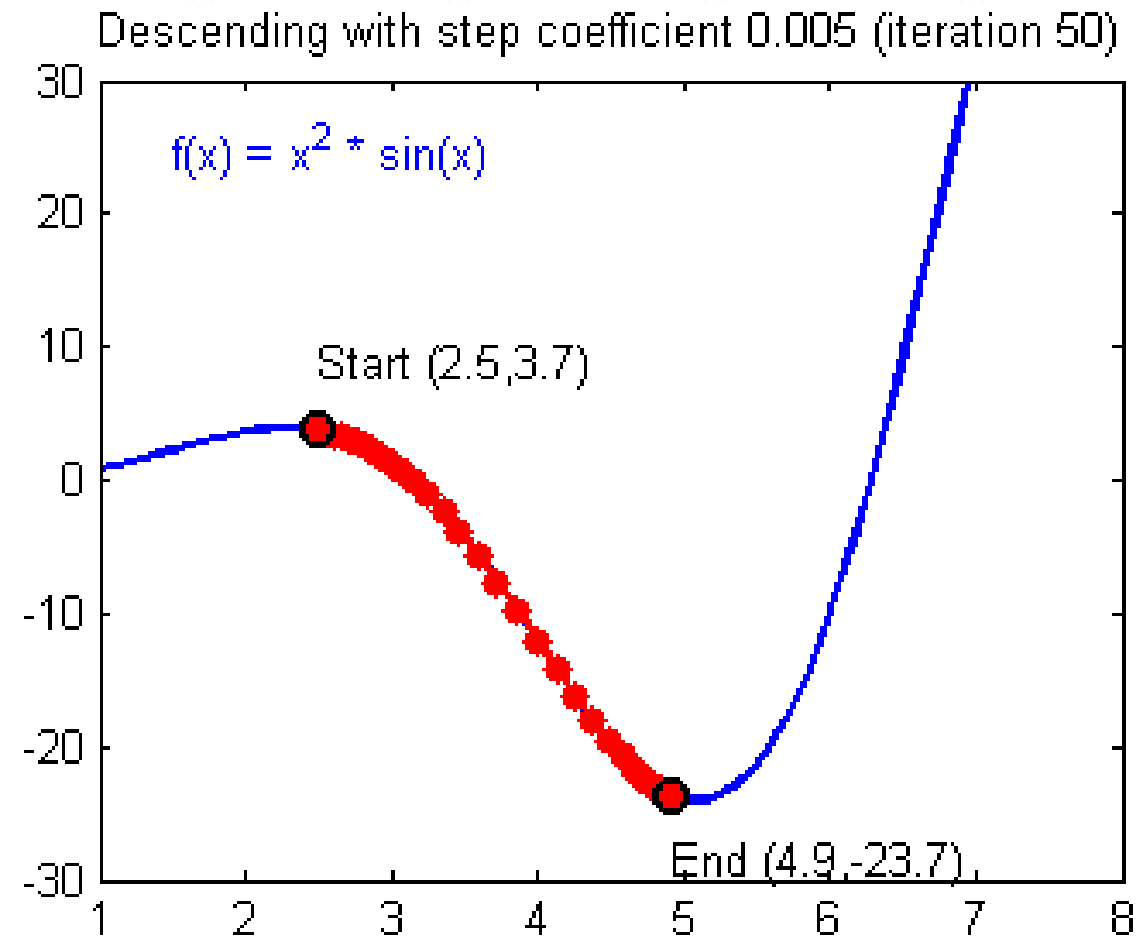
$$\begin{aligned}\theta_1 &:= \theta_1 - \alpha \frac{d}{d\theta_1} J(\theta_1, \theta_2) \\ \theta_2 &:= \theta_2 - \alpha \frac{d}{d\theta_2} J(\theta_1, \theta_2)\end{aligned}$$

Derivatives:

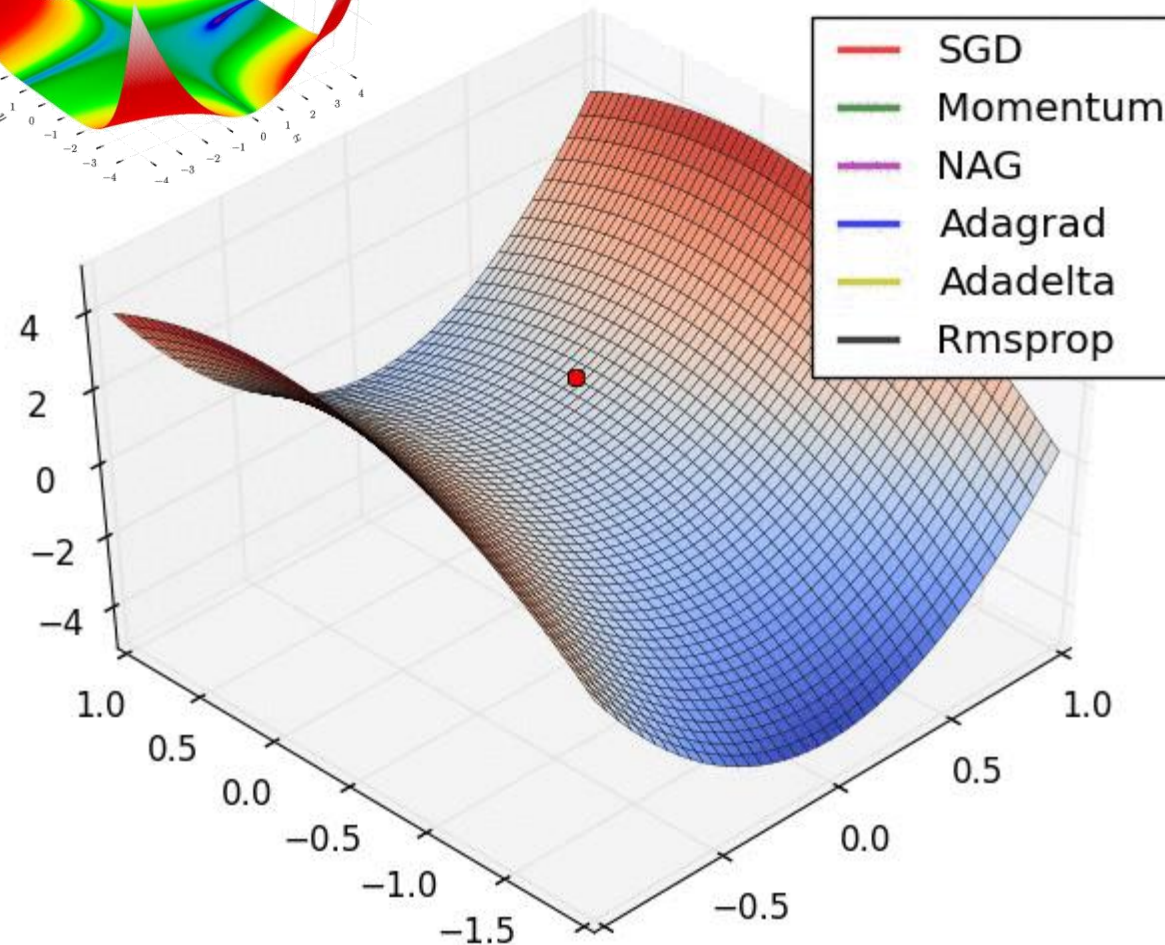
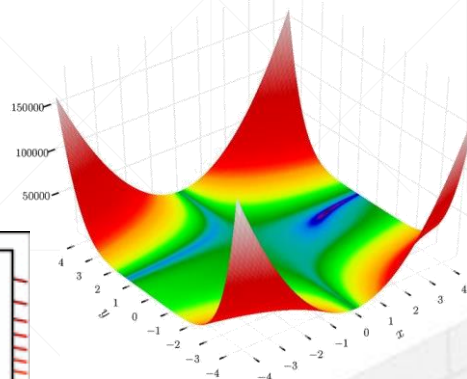
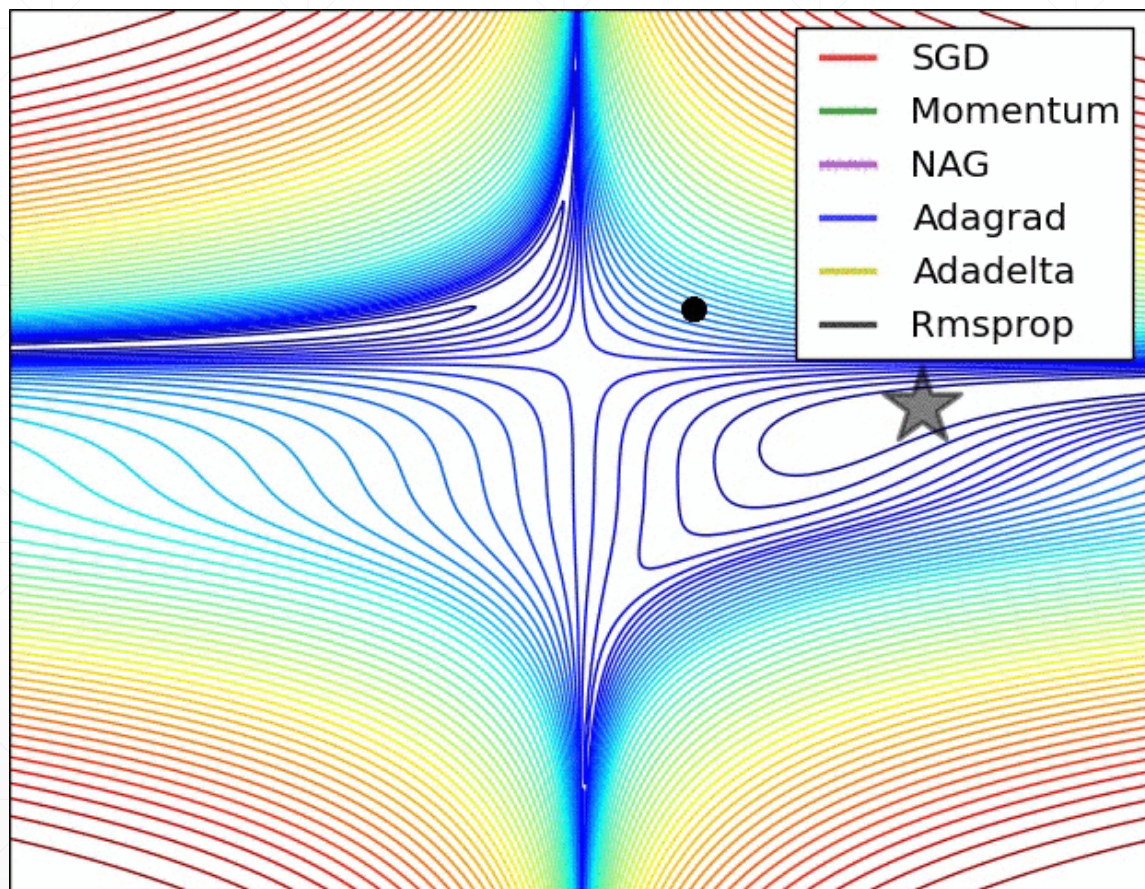
$$\frac{d}{d\theta_1} J(\theta_1, \theta_2) = \frac{d}{d\theta_1} \theta_1^2 + \frac{d}{d\theta_1} \theta_2^2 = 2\theta_1$$

$$\frac{d}{d\theta_2} J(\theta_1, \theta_2) = \frac{d}{d\theta_2} \theta_1^2 + \frac{d}{d\theta_2} \theta_2^2 = 2\theta_2$$

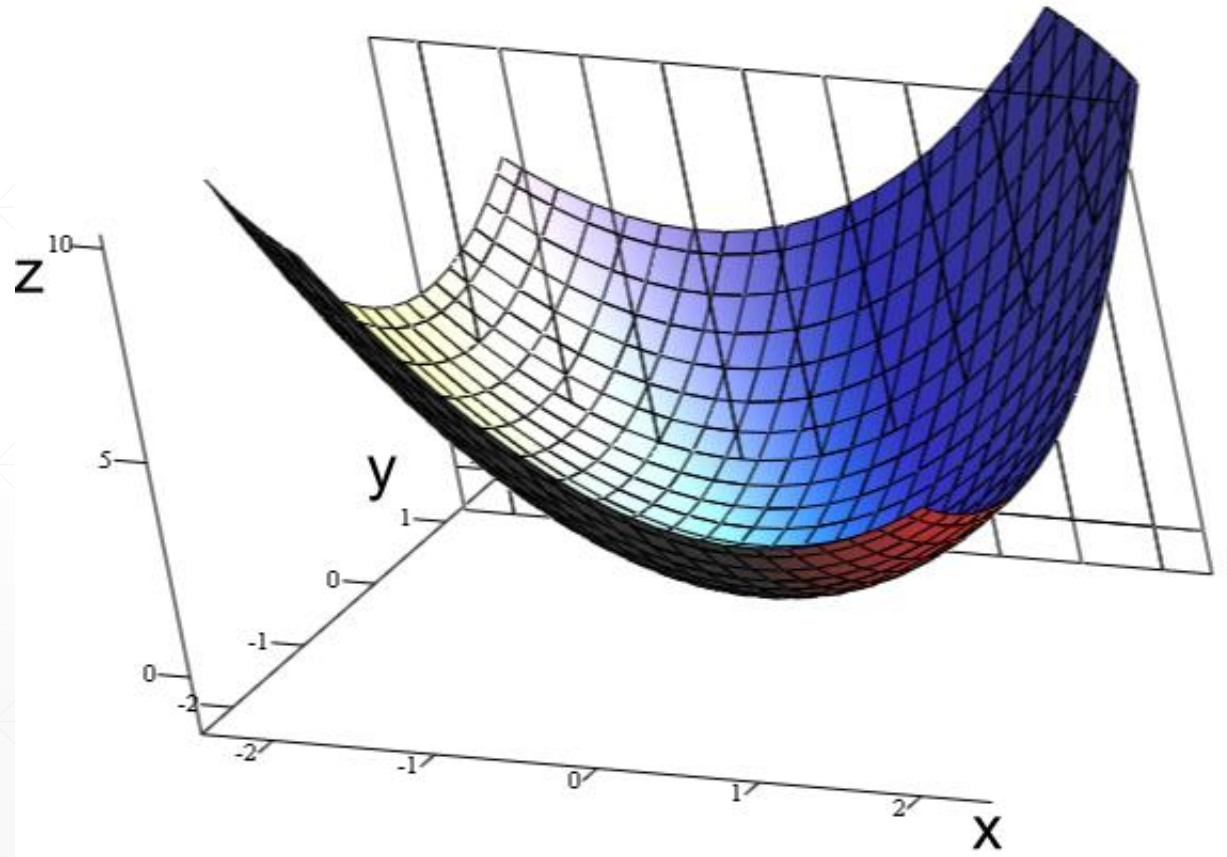
Learning process-1



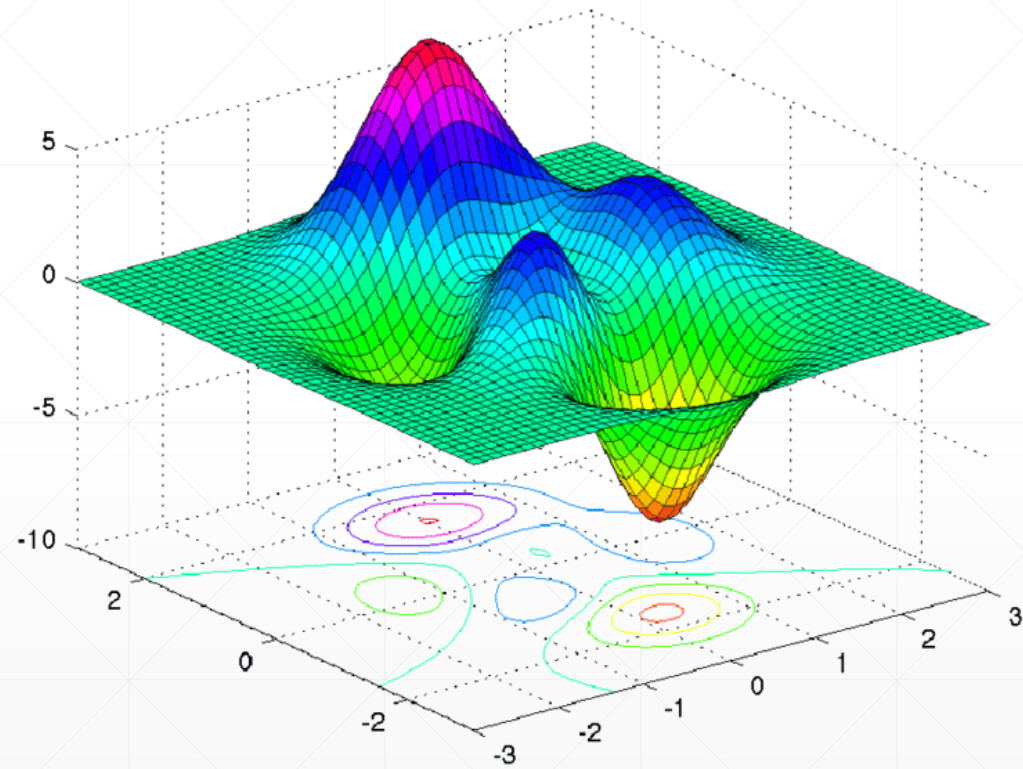
Learning process-2



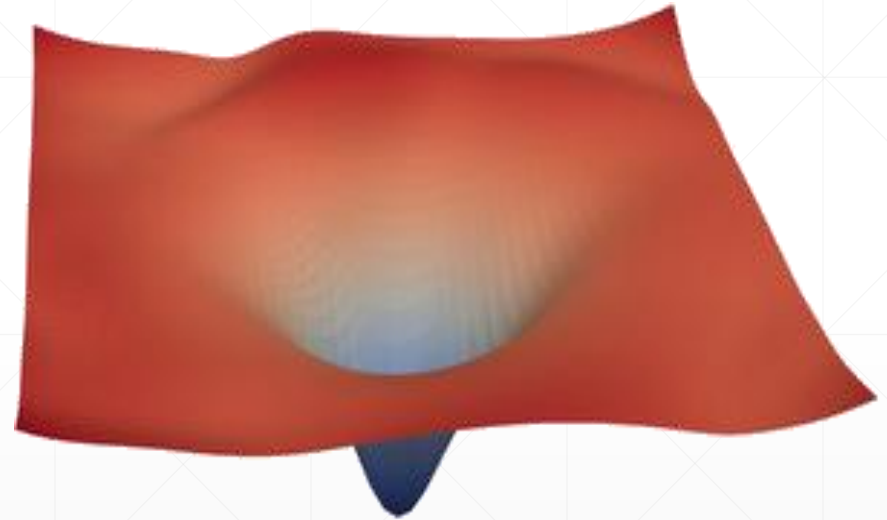
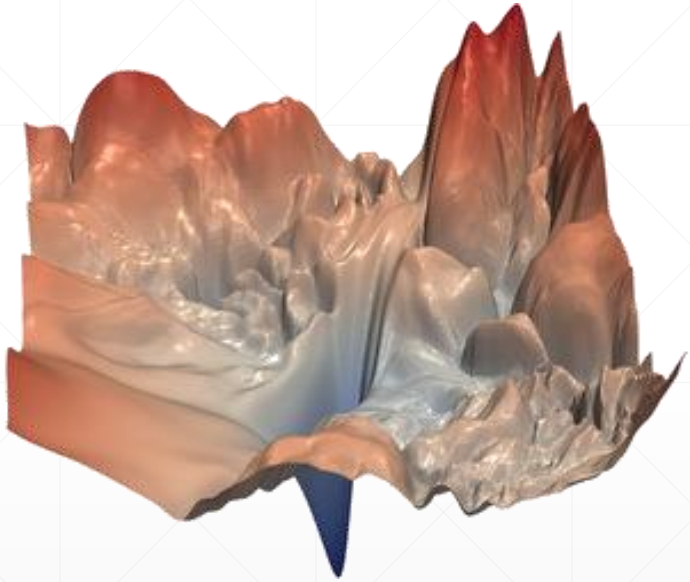
Convex function



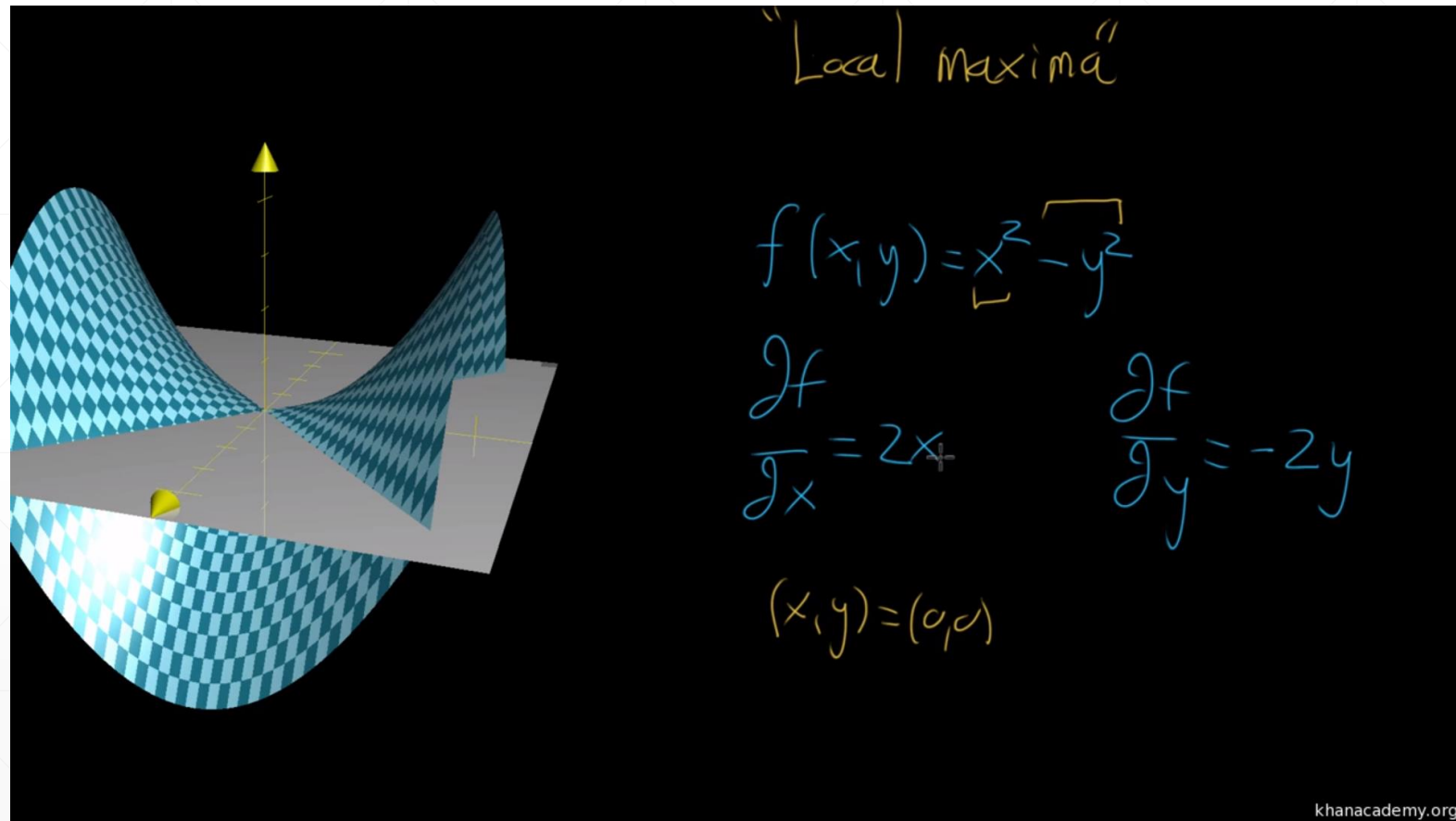
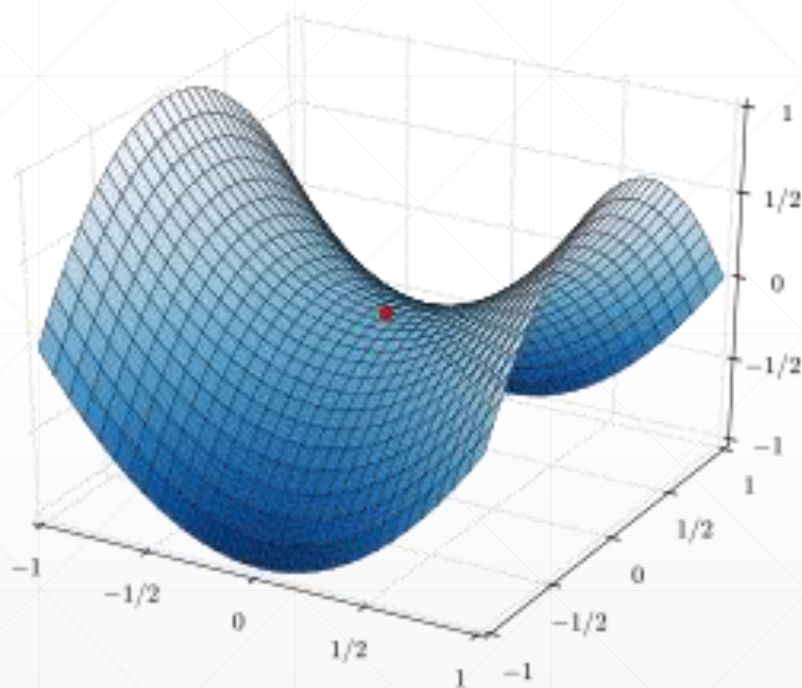
Local Minima



ResNet-56



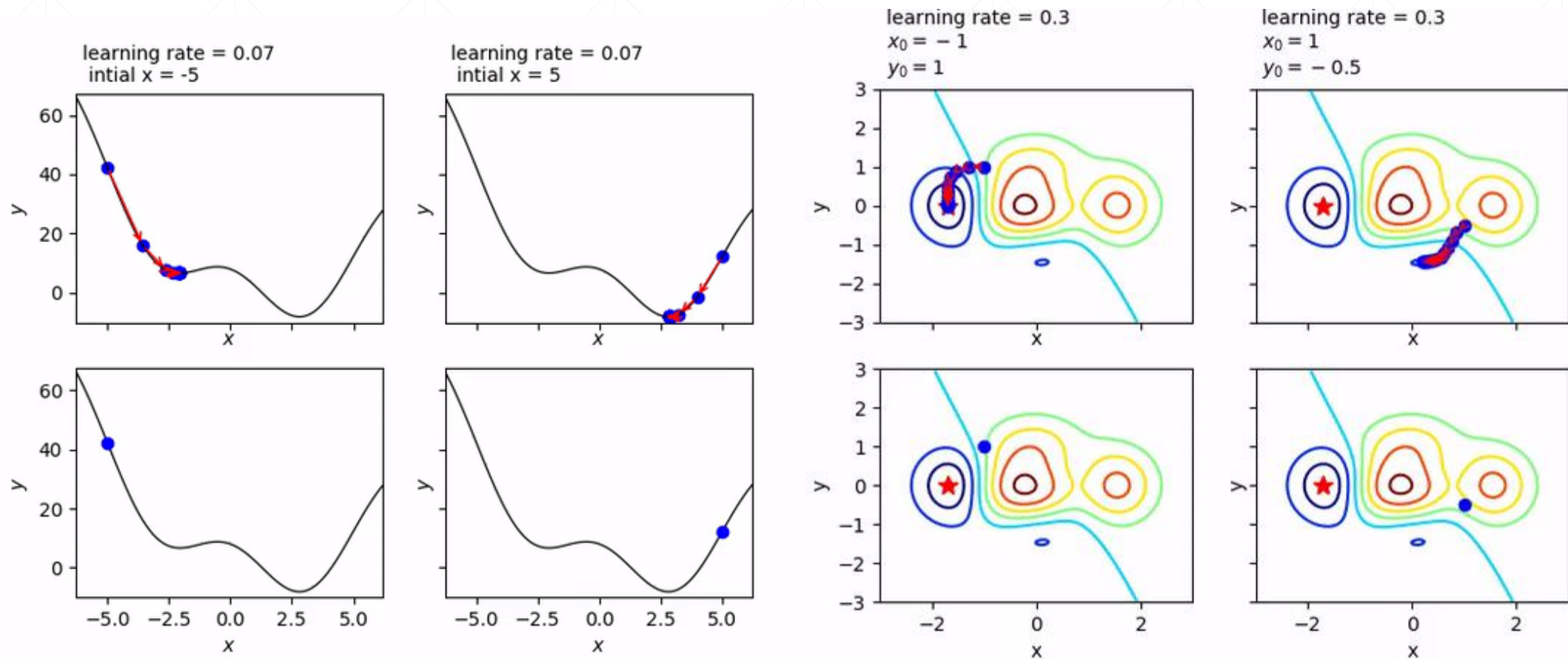
Saddle point



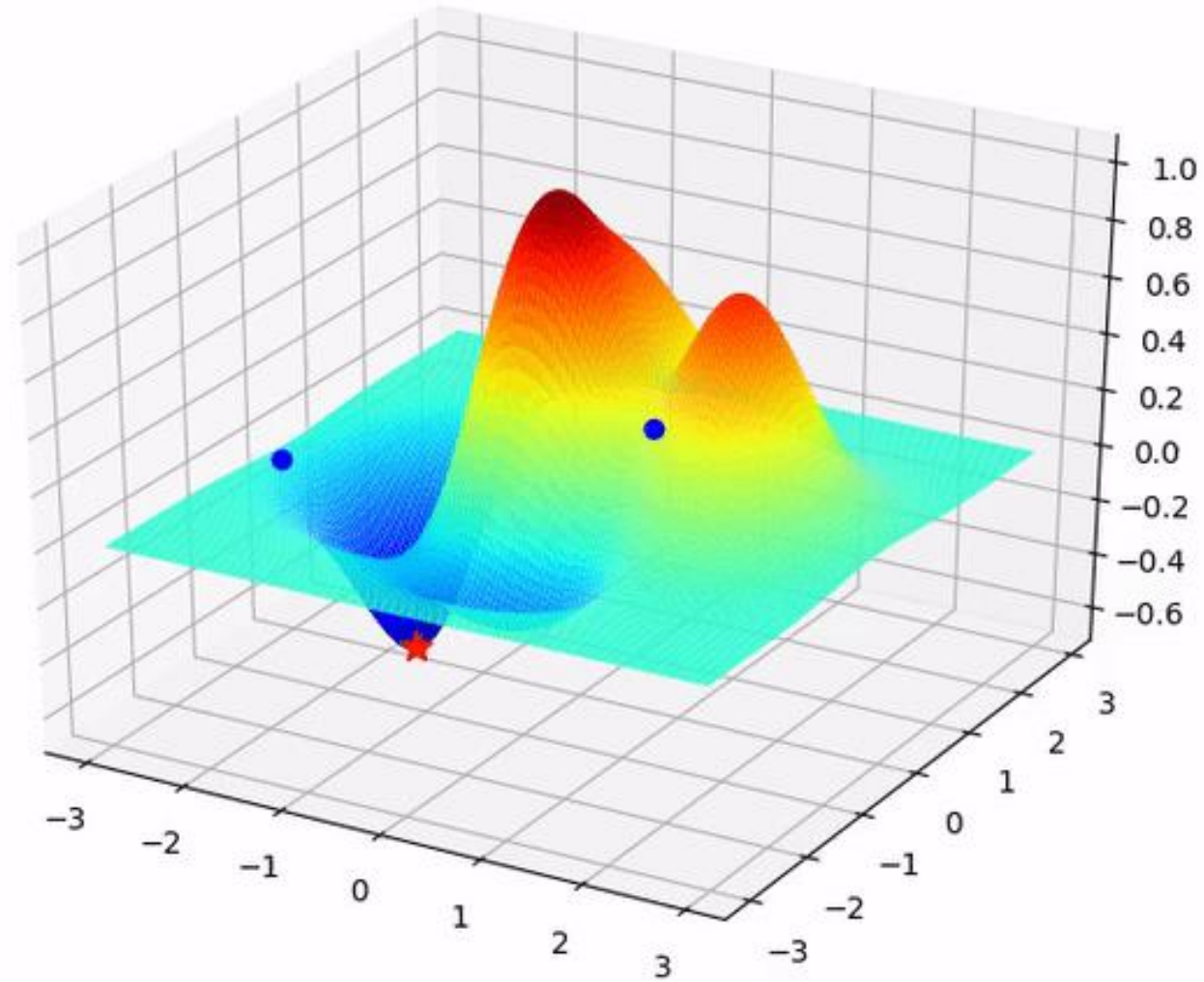
Optimizer Performance

- initialization status
 - learning rate
 - momentum
 - etc.
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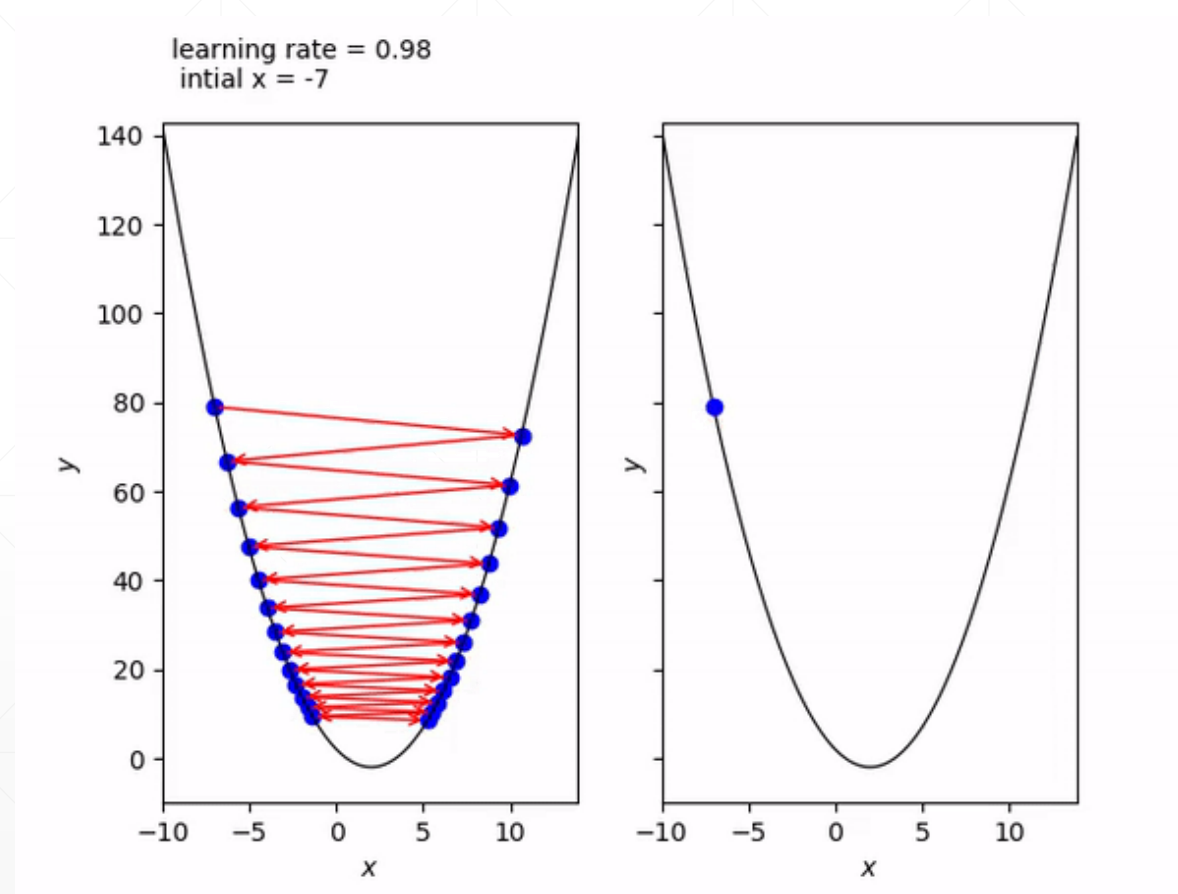
Initialization



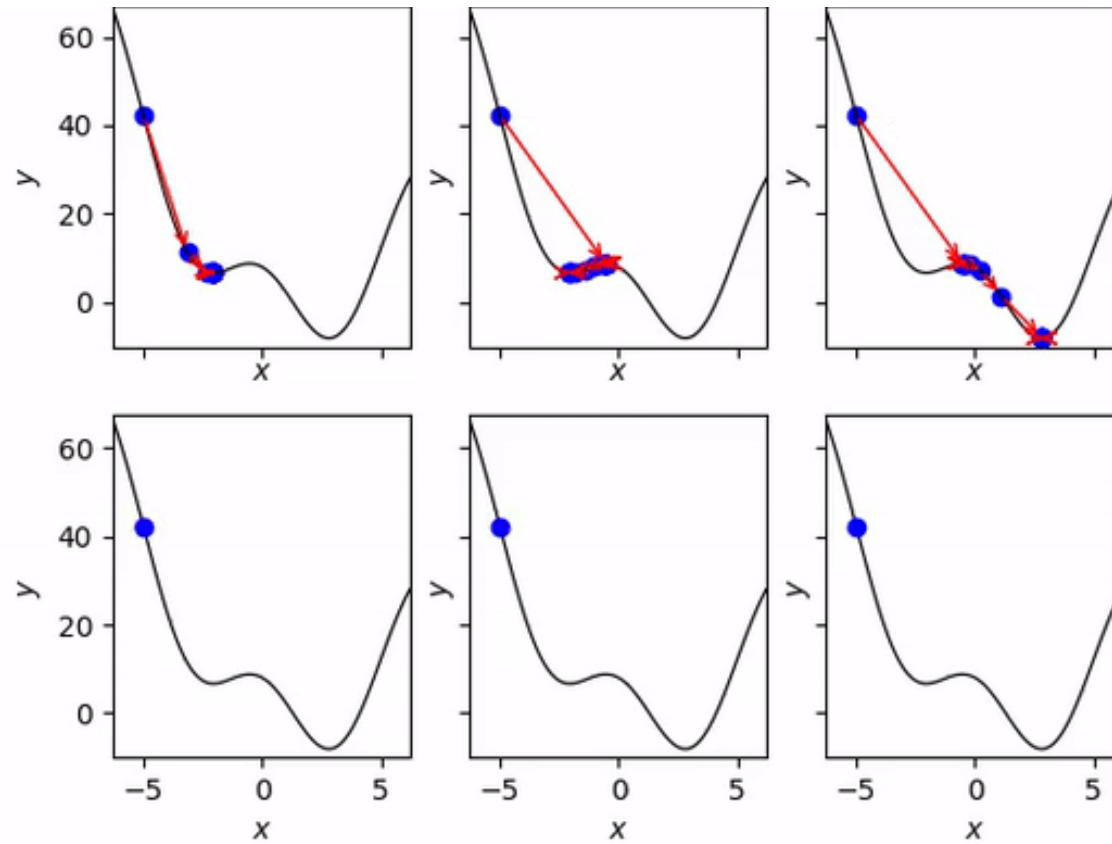
Initialization



Learning rate



Escape minima



下一课时

常见函数梯度

Thank You.
