



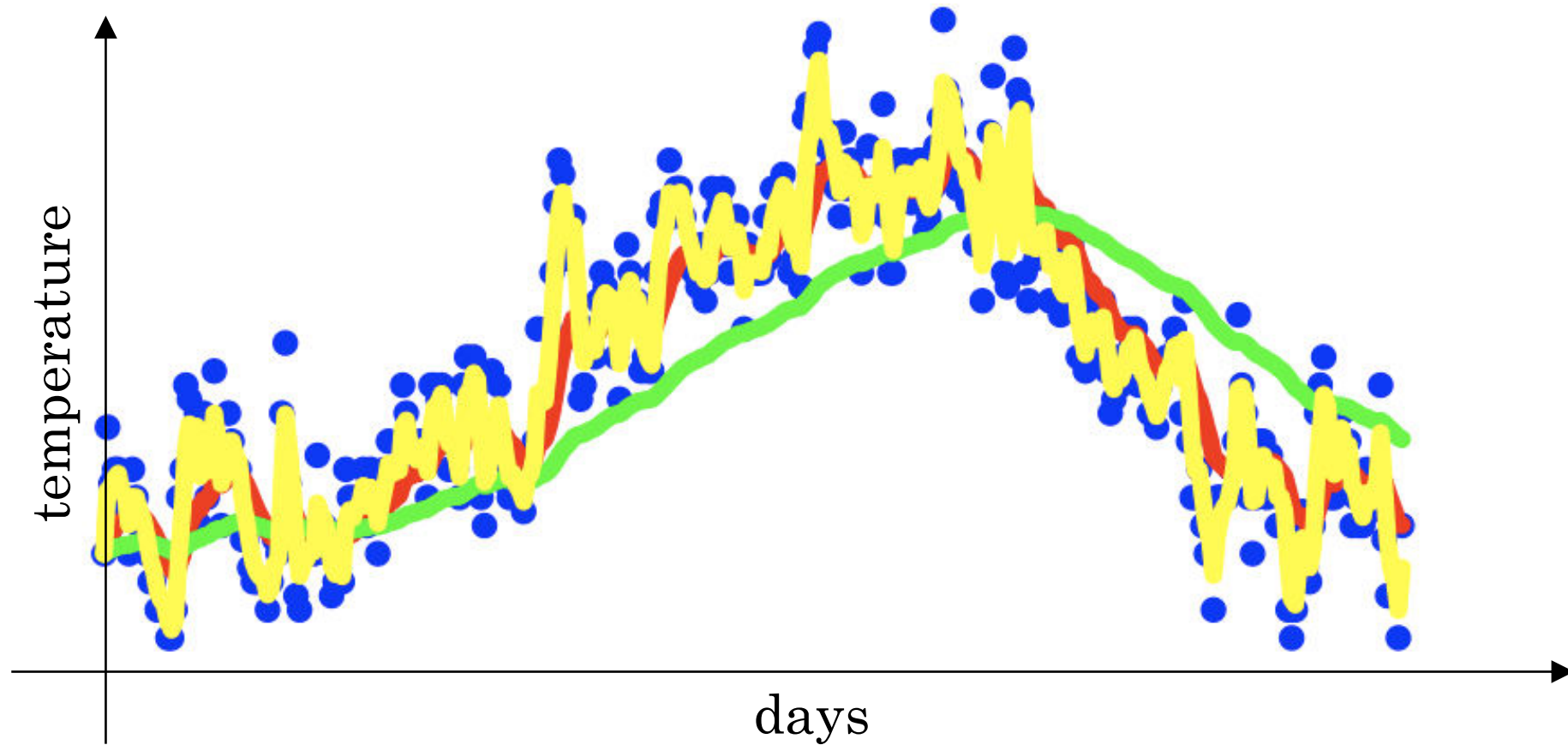
deeplearning.ai

Optimization Algorithms

Understanding
exponentially
weighted averages

Exponentially weighted averages

$$v_t = \beta v_{t-1} + (1 - \beta)\theta_t$$



Implementing exponentially weighted averages

$$v_0 = 0$$

$$v_1 = \beta v_0 + (1 - \beta) \theta_1$$

$$v_2 = \beta v_1 + (1 - \beta) \theta_2$$

$$v_3 = \beta v_2 + (1 - \beta) \theta_3$$

...

$$V_\theta = 0$$

$$V_\theta = \beta V + (1 - \beta) \theta_1$$

$$V_\theta = \beta V + (1 - \beta) \theta_2$$

$$V_\theta = \beta V + (1 - \beta) \theta_3$$

...

⇒ we don't have to maintain $V_1, V_2, V_3 \dots$

$$V_\theta = 0$$

Repeat {

Get θ for next day

$$V_\theta = \beta V_\theta + (1 - \beta) \theta_t$$

}

low
memory but
not the best
way to compute
moving avg