

# Learning Rate Optimizing

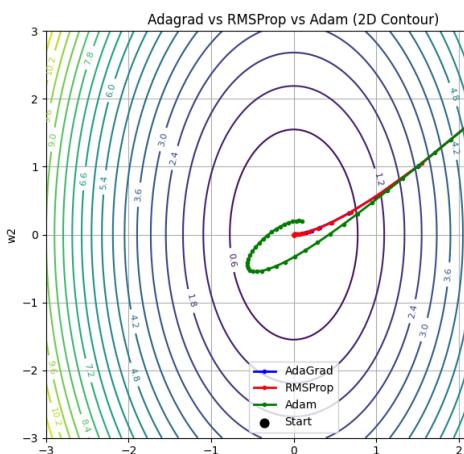
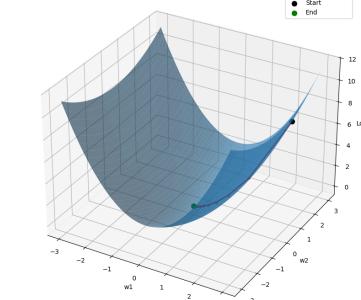
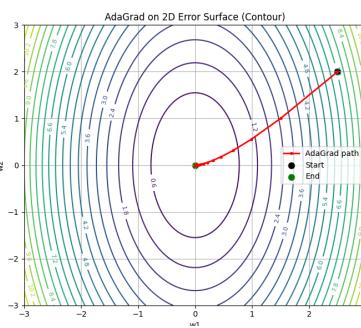
## AdaGrad

**Rhyme**  
 “Square the slope and make a stack,  
 Bigger the pile, the learning steps  
 slack.  
 Divide by the root so updates shrink  
 —  
 AdaGrad slows with every blink.”

- ✓ “Square the slope” →  $(\nabla_\theta)^2$
- ✓ “make a stack” → cumulative  $G = G + \dots$
- ✓ “updates shrink” → denominator grows
- ✓ “slows with every blink” → learning rate decays

$$G = G + (\nabla_\theta)^2$$

$$\theta = \theta - \eta \frac{\nabla_\theta}{\sqrt{G} + \epsilon}$$



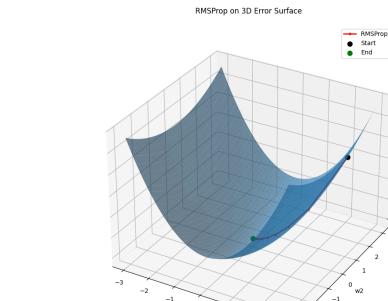
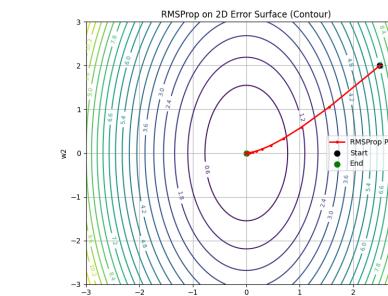
## RMSProp

**Rhyme**  
 “Forget the past but not all the way,  
 Keep a soft memory with beta’s sway.  
 Smooth the squares and divide to drop  
 —  
 That’s the rhythm of RMSProp.”

- ✓ “soft memory” → moving average
- ✓ “beta’s sway” →  $\beta$  weighting
- ✓ “smooth the squares” → averaging squared gradients
- ✓ “divide to drop” → update rule denominator

$$v_t = \beta * v_{t-1} + (1 - \beta)(\nabla w_t)^2$$

$$w_{t+1} = w_t - \frac{\eta}{\sqrt{v_t + \epsilon}} * \nabla w_t$$



## Adam

**Rhyme**  
 “First the mean and second the  
 square,  
 Moments two dance in Adam’s lair.  
 Bias corrected, the steps align —  
 Adaptive moves that learn just  
 fine.”

- ✓ “mean” → first moment  $m$
- ✓ “square” → second moment  $v$
- ✓ “bias corrected” →  $\hat{m}$  and  $\hat{v}$
- ✓ “adaptive moves” → denominator adjusts per parameter

$$m = \beta_1 m + (1 - \beta_1)\nabla_\theta$$

$$v = \beta_2 v + (1 - \beta_2)\nabla_\theta^2$$

$$\hat{m} = \frac{m}{1 - \beta_1^t}, \quad \hat{v} = \frac{v}{1 - \beta_2^t}$$

$$\theta = \theta - \eta \frac{\hat{m}}{\sqrt{\hat{v}} + \epsilon}$$

