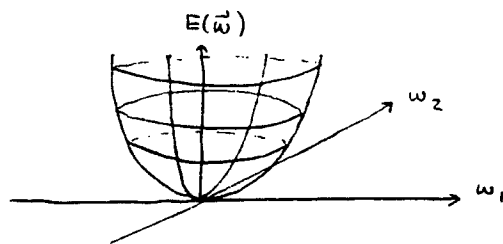


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## Gradient Descent - Quadratic Bowls

claim: Near a minimum, every infinitely differentiable function of  $n$  variables is approximately a quadratic bowl.



Quadratic bowl for 2 variables. Bottom of bowl need not be at height = 0. Need not be at  $w_1 = w_2 = 0$ , etc.

pf: Let  $\vec{w}_0$  be at a minimum of  $E(\vec{w})$

Consider a small perturbation of  $\vec{w}_0$ :

$$\vec{w} = \vec{w}_0 + \vec{\epsilon} \quad \text{where } \vec{\epsilon} \text{ is small vector}$$

Then by Taylor series:

$$E(\vec{w}) = E(\vec{w}_0 + \vec{\epsilon}) = E(\vec{w}_0) + \nabla_{\vec{w}} E(\vec{w}) \Big|_{\vec{w}=\vec{w}_0} \cdot \vec{\epsilon} + \frac{\vec{\epsilon}^T H \vec{\epsilon}}{2} + O(|\vec{\epsilon}|^3)$$

since  $O(|\epsilon|^3)$  small "  $\approx E(\vec{w}_0) + \nabla_{\vec{w}} E(\vec{w}) \Big|_{\vec{w}=\vec{w}_0} \cdot \vec{\epsilon} + \frac{\vec{\epsilon}^T H \vec{\epsilon}}{2}$

But  $\nabla_{\vec{w}} E(\vec{w}) \Big|_{\vec{w}=\vec{w}_0} = 0$  since  $E(\vec{w})$  is flat at bottom of bowl.

So we have  $E(\vec{w}_0 + \vec{\epsilon}) = E(\vec{w}_0) + \frac{\vec{\epsilon}^T H \vec{\epsilon}}{2}$   
const quadratic

i.e. we have quadratic bowl.

Note:  $H$  is Hessian  $\equiv \left[ \frac{\partial^2 E(\vec{w})}{\partial w_j \partial w_i} \right]_{\vec{w}=\vec{w}_0}$   
(matrix)