

Homework (3)

1. Compute the gradient and Hessian of the following functions:

a. $f(\mathbf{x}) = \|\mathbf{x}\|_p, \mathbf{x} \neq \mathbf{0}, p > 1.$

b. $f(\mathbf{x}) = (\mathbf{a}^T \mathbf{x})(\mathbf{b}^T \mathbf{x}).$

c. $f(\mathbf{x}) = \frac{1}{2} \|\mathbf{Ax} - \mathbf{b}\|_2^2.$

Homework (3)

2. Compute the gradient of the following matrix functions (write in matrices, rather than entrywise. Give details. Working out 5 is sufficient if you don't want to torture yourself.):

a. $f(\mathbf{X}) = \|\mathbf{X}\|_F.$

b. $f(\mathbf{X}) = \|\mathbf{X}^T \mathbf{A} \mathbf{X}\|_F^2$ and \mathbf{A} is a symmetric matrix.

c. $f(\mathbf{X}) = \|\text{diag}(\mathbf{X}^T \mathbf{A} \mathbf{X})\|_F^2$ and \mathbf{A} is a symmetric matrix.

d. $f(\mathbf{X}) = \det \mathbf{X}.$

e. $f(\mathbf{X}) = \det(\mathbf{X}^T \mathbf{A} \mathbf{X}).$

f. $f(\mathbf{X}) = \mathbf{a}^T \mathbf{X}^{-1} \mathbf{b}.$

g. $f(\mathbf{X}) = \text{tr}(\mathbf{A} \mathbf{X} \mathbf{B}).$

h. $f(\mathbf{X}) = \text{tr}(\mathbf{A} \mathbf{X}^{-1} \mathbf{B}).$

i. $f(\mathbf{X}) = \text{tr}((\mathbf{A} + \mathbf{X})^{-1}).$

Homework (3)

3. Write a program for computing the gradient of error function V with respect to the weights of a neural network with arbitrary topology. Test it on simplified ResNet: 1 filter, 1 skip connection, no pooling and batch normalization. Verify the correctness by numerical differentiation, i.e. verify

$$\frac{V(\mathbf{w} + t\mathbf{v}) - V(\mathbf{w})}{t} \approx \langle \nabla V(\mathbf{w}), \mathbf{v} \rangle$$

for arbitrary choice of \mathbf{v} and sufficiently small t . The code should include the verification step.

4. Write an automatic differentiation program that works on a given expression DAG. Test it on function y_o and verify it with numerical differentiation:

$$y_o = (\sin(x_1+1)+\cos(2x_2)) \tan(\log(x_3))+(\sin(x_2+1)+\cos(2x_1)) \exp(1+\sin(x_3)).$$

The code should include the verification step.

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5. Find the dual norm of Mahalanobis norm: $\|\mathbf{x}\|_{\mathbf{M}} = \sqrt{\mathbf{x}^T \mathbf{M} \mathbf{x}}$, where \mathbf{M} is a positive definite matrix.
6. Prove that the eigenvalues λ_i of $(\mathbf{A} + \mathbf{B})^{-1} \mathbf{A}$, where \mathbf{A} is positive semidefinite and \mathbf{B} is positive definite, satisfy $0 \leq \lambda_i < 1$.
7. Compute the condition number of the following matrix:

$$\begin{bmatrix} 1 & 2 & 3 \\ 3 & 4 & 5 \\ 5 & 6 & 9 \end{bmatrix}.$$

8. Suppose $\mathbf{X} \in \mathbb{R}^{3 \times 3}$, $\mathcal{A}(\mathbf{X}) = X_{11} + X_{12} - X_{31} + 2X_{33}$, find \mathcal{A}^* .