

Complex Manifolds training loop, used for Bayesian networks.

This document is intended to explain exactly how the contour integration loop I am working on, which is designed to work with Bayesian networks to reduce the training time and improve the accuracy involve in these, actually works and why I am using the particular methods that I am doing to implement this.

What does it do?

It calculates partial derivatives in

\mathbb{C}^n , where n can be as many dimensions as required.

How does it work?

This algorithm works by using a Logarithmic derivative lemma, which unlike alternatives such as the Wirtinger derivative does not require a closed form for a derivative.

For example a closed form derivative may look like this:

$\frac{\partial}{\partial x}(x^2 y) = 2xy$, where as the one produced
from this algorithm may be more numerical.

A Wirtinger derivative is more like the above, where as the algorithm I am creating relies on using contour integration around a branch cut to do it. for example:

$$\oint_C \frac{1}{2} \cdot \text{Log} \left(\sum_{j=0}^n |\psi_j|^2 \right) - \frac{1}{2} \text{Log} \left(\sum_{i=k}^n |\psi_j(0)|^2 \right) dt, \text{ where } t \text{ is the parameter used to}$$

do this integral **and** C is a smooth path around the branch cut,
taken along the negative real axis.

The argument in this case is defined as the principal argument,
i.e. -

$\pi \leq \arg z < 2\pi$, although this can be done with other argument values **to** get different results,
depending on what result is desired • li

References:

M.Schneider, Yum-Tong Siu, Several Complex Variables (37).

https://en.wikipedia.org/wiki/Wirtinger_derivatives [online] (accessed 03/06/2022)

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