# On Stability of Newton Schulz Iterations in an Approximate Algebra

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#### I. INTRODUCTION

In many disciplines, finite correlations coorespond to matrices with decay properties. Matrix decay involves an approximate (perhaps bounded) inverse relationship between matrix elements and a cooresponding distance; this may be a simple inverse exponential relationship between elements and the Cartesian distance between support functions, or it may involve a statistical distance, e.g. between strings. In electronic structure, correlations manifest in decay properties of the matrix sign function, as projector of the effective Hamiltonian. More broadly, matrix decay properties may coorespond to learned correlations in a generalized, non-orthogonal metric, obtained perhaps through first order optimization involving the celebrated PLSEV line search approach to semi-definate programming based on the matrix sign function. More broadly still, problems with local, non-orothogonal support are often solved with congruential transformations based on the matrix inverse square root or inverse factor; these transformations correlate the non-orthogonal support in a representation independent form, eq. of the generalized eigenproblem.

These problems are related through Higham's identity, connecting the matrix sign function with the inverse square root:

Computation of these matrix functions may encounter ill-conditioning, cooresponding to extended (quasi-degenerate) correlations and near complete non-orthogonal support and associated slow rates of decay. For extremely slow decay, maybe even oscillatory, low order algebraic decay, methods that compresion.... For fast decay,

correlation and the support

Also, matrices with decay arise from the application \*Electronic address: matt.challacombe@freeon.org; URL: http://www.freeon.org

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of . Generally, ill-conditining is associated with slower decay,

Decay principles, often very sparse but very illconditioned problems.

#### A. Incomplete/Inexact Schemes

Incompleteness -; sparse approximations dense problems, uses conventional sparse infrastructure, second order errors in matrix multiplication. Often adhoc.

$$\begin{split} \tilde{\pmb{a}} &= \pmb{a} + \delta \pmb{a} \\ \tilde{\pmb{a}} \cdot \tilde{\pmb{b}} &= \pmb{a} \cdot \pmb{b} + \delta \pmb{a} \cdot \pmb{b} + \pmb{a} \cdot \delta \pmb{b} + \delta \pmb{a} \cdot \delta \pmb{b} \end{split}$$

The variations do not express in the overal context of the product. Because the error in the incomplete case is additive, the

For example, a may be small, but  $\delta a \cdot b$  large, leading extra work. Also, once a truncation error is committed, it is encounted in all subsiquent steps; it becomes difficult to impossible to manage error flows of differing magnitude in complex maps.

#### B. Retaining the Eigenspace

Gradients lack convergence properties Iteration without orig drives away from basis NS has both. Difference between scalar iteration, Higham page 92.

# C. Approximate Algebra as N-Body Problem

SpAMM is the recursive Cauchy-Schwarz occlusion product  $\otimes_{\tau}$  on matrix quadtrees

$$\boldsymbol{a}^{i} = \begin{bmatrix} \boldsymbol{a}_{00}^{i+1} & \boldsymbol{a}_{01}^{i+1} \\ \boldsymbol{a}_{10}^{i+1} & \boldsymbol{a}_{11}^{i+1} \end{bmatrix}$$
 (2)

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$$\boldsymbol{a}^{i} \otimes_{\tau} \boldsymbol{b}^{i} = \begin{cases} \emptyset & \text{if } \|\boldsymbol{a}^{i}\| \|\boldsymbol{b}^{i}\| < \tau \\ \boldsymbol{a}^{i} \cdot \boldsymbol{b}^{i} & \text{if}(i = \text{leaf}) \\ \left[\boldsymbol{a}_{00}^{i+1} \otimes_{\tau} \boldsymbol{b}_{00}^{i+1} + \boldsymbol{a}_{01}^{i+1} \otimes_{\tau} \boldsymbol{b}_{10}^{i+1} , & \boldsymbol{a}_{00}^{i+1} \otimes_{\tau} \boldsymbol{b}_{01}^{i+1} + \boldsymbol{a}_{01}^{i+1} \otimes_{\tau} \boldsymbol{b}_{11}^{i+1} \right] \\ \boldsymbol{a}_{00}^{i+1} \otimes_{\tau} \boldsymbol{b}_{01}^{i+1} + \boldsymbol{a}_{01}^{i+1} \otimes_{\tau} \boldsymbol{b}_{11}^{i+1} , & \boldsymbol{a}_{00}^{i+1} \otimes_{\tau} \boldsymbol{b}_{01}^{i+1} + \boldsymbol{a}_{01}^{i+1} \otimes_{\tau} \boldsymbol{b}_{11}^{i+1} \right] \end{cases} \text{ else}$$

database orientation, Cauchy sch Approximate Algebra, SpAMM Cauchy Schwarz occlusion, n-body approach to numerical linear algebra, first order errors in matrix multiplication. Based on Cauchy Schwarz inequality.

$$\mathbf{a} \otimes_{\tau} \mathbf{b} = \mathbf{a} \cdot \mathbf{b} + \mathbf{\Delta}_{\tau}^{a \cdot b} \tag{4}$$

where  $\Delta_{\tau}^{a \cdot b}$  is a deterministic (assymetric) first order variation cooresponding to the branch pattern set by Cauchy-Schwarz occlusion, with length  $\|\Delta_{\tau}^{a \cdot b}\| \leq \tau \|a\| \|b\|$ . The opperator  $\otimes_{\tau}$  leads to a non-associative algebra with Lie bracket

$$[\boldsymbol{a}, \boldsymbol{b}]_{\tau} = \boldsymbol{a} \otimes_{\tau} \boldsymbol{b} - \boldsymbol{b} \otimes_{\tau} \boldsymbol{a} = [\boldsymbol{a}, \boldsymbol{b}] + \boldsymbol{\Delta}_{\tau}^{a \cdot b} - \boldsymbol{\Delta}_{\tau}^{b \cdot a}.$$
 (5)

determined by the occlusion field. Our challenge is to master the error flows of these occlusion fields under iteration, for ill-conditioned problems and with permisive values of  $\tau$ .

#### II. NEWTON SHULZ ITERATION

#### A. Idempotence

#### B. The Scaled Map

### C. Alternative Formulations

dual, stabilized and naive

## III. OCCLUSION FLOWS

 $\delta x_k$  and  $\delta z_k$  arrize from itteration with  $\otimes_{\tau}$ , and are deterministic flows away from the manifold of s determined by sensitivity of the NS iteration to these numerical insults.

$$\delta \boldsymbol{x}_{k}^{\text{naiv}} = \delta \widetilde{\boldsymbol{z}}_{k} \cdot \boldsymbol{s} \cdot \widetilde{\boldsymbol{z}}_{k} + \widetilde{\boldsymbol{z}}_{k} \cdot \boldsymbol{s} \cdot \delta \widetilde{\boldsymbol{z}}_{k}$$
 (6)

$$\delta \boldsymbol{x}_{k}^{\text{dual}} = \delta \widetilde{\boldsymbol{y}}_{k} \cdot \widehat{\boldsymbol{z}}_{k} + \widetilde{\boldsymbol{y}}_{k} \cdot \delta \widetilde{\boldsymbol{z}}_{k}$$
 (7)

$$\widetilde{\boldsymbol{x}}_{k} = f\left[\widetilde{\boldsymbol{z}}_{k-1}, \widetilde{\boldsymbol{x}}_{k-1}\right] 
= \mathbf{m}\left[\widetilde{\boldsymbol{x}}_{k-1}\right] \cdot \widetilde{\boldsymbol{z}}_{k-1}^{\dagger} \cdot \boldsymbol{s} \cdot \widetilde{\boldsymbol{z}}_{k-1} \cdot \mathbf{m}\left[\widetilde{\boldsymbol{x}}_{k-1}\right]$$
(8)

$$\delta x_k = f_{\delta z_{k-1}} \|\delta z_{k-1}\| + f_{\delta x_{k-1}} \|\delta x_{k-1}\| + \mathcal{O}(\tau^2)$$
 (9)

generalized Gateaux differential

$$f_{\delta \boldsymbol{z}_{k-1}} = \lim_{\tau \to 0} \frac{f[\boldsymbol{z}_{k-1} + \tau \delta \widehat{\boldsymbol{z}}_{k-1}, \widetilde{\boldsymbol{x}}_{k-1}] - f[\boldsymbol{z}_{k-1}, \widetilde{\boldsymbol{x}}_{k-1}]}{\tau}$$
$$= L_{\widetilde{\boldsymbol{x}}_{k}} (\widetilde{\boldsymbol{z}}_{k}, \delta \widehat{\boldsymbol{z}}_{k-1})$$
(10)

$$f_{\delta \boldsymbol{x}_{k-1}} = \lim_{\tau \to 0} \frac{f[\widetilde{\boldsymbol{z}}_{k-1}, \boldsymbol{x}_{k-1} + \tau \delta \widehat{\boldsymbol{x}}_{k-1}] - f[\widetilde{\boldsymbol{z}}_{k-1}, \boldsymbol{x}_{k-1}]}{\tau}$$
$$= L_{\widetilde{\boldsymbol{x}}_{k}}(\widetilde{\boldsymbol{z}}_{k}, \delta \widehat{\boldsymbol{x}}_{k-1})$$
(11)

$$L_{\widetilde{\boldsymbol{x}}_{k}}\left(\widetilde{\boldsymbol{z}}_{k}, \delta\widehat{\boldsymbol{x}}_{k-1}\right) = \delta\widehat{\boldsymbol{x}}_{k-1}^{\dagger} \cdot \mathbf{m}'\left[\boldsymbol{x}_{k-1}\right] \cdot \left\{\widetilde{\boldsymbol{z}}_{k-1}^{\dagger} \cdot \boldsymbol{s} \cdot \widetilde{\boldsymbol{z}}_{k}\right\} + \left\{\widetilde{\boldsymbol{z}}_{k}^{\dagger} \cdot \boldsymbol{s} \cdot \widetilde{\boldsymbol{z}}_{k-1}\right\} \cdot \mathbf{m}'\left[\boldsymbol{x}_{k-1}\right] \cdot \delta\widehat{\boldsymbol{x}}_{k-1} \quad (12)$$

$$L_{\widetilde{\boldsymbol{x}}_{k}}\left(\widetilde{\boldsymbol{z}}_{k}, \delta\widehat{\boldsymbol{z}}_{k-1}\right) = \left\{ \mathbf{m} \left[\boldsymbol{x}_{k-1}\right] \cdot \delta\widehat{\boldsymbol{z}}_{k-1}^{\dagger} \cdot \boldsymbol{s} \right\} \cdot \widetilde{\boldsymbol{z}}_{k} + \widetilde{\boldsymbol{z}}_{k}^{\dagger} \cdot \left\{ \boldsymbol{s} \cdot \delta\widehat{\boldsymbol{z}}_{k-1} \cdot \mathbf{m} \left[\boldsymbol{x}_{k-1}\right] \right\} \quad (13)$$

$$\{\widetilde{\boldsymbol{z}}_{k}^{\dagger} \cdot \boldsymbol{s} \cdot \widetilde{\boldsymbol{z}}_{k-1}\} \to \boldsymbol{p}_{+}[\boldsymbol{s}]$$
 (14)

$$\{s \cdot \delta \widehat{\boldsymbol{z}}_{k-1} \cdot \mathbf{m} [\boldsymbol{x}_{k-1}]\} \to \boldsymbol{n} [s]$$
 (15)

# IV. BASIS SET ILL-CONDITIONING IN ELECTRONIC STRUCTURE

A. 3,3 carbon nanotube with diffuse sp-function double exponential (Fig.)

# B. Water with triple zeta and double polarization

Here's looking at you Jurg...

## V. IMPLEMENTATION

# A. Methods

FP, F08, OpenMP 4.0

### B. A Modified NS Map

C.  $\delta \boldsymbol{x}_k$  and  $\delta \boldsymbol{x}_k$  channels

tau= Figure showing channels etc.

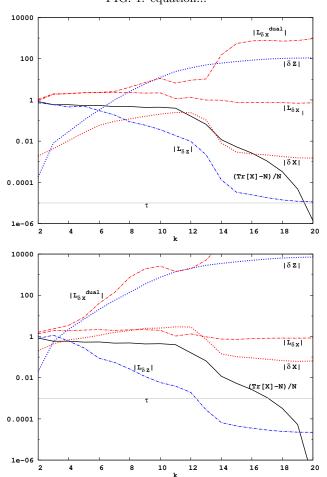
# D. Convergence

Map switching and etc based on TrX

# VI. EXPERIMENTS

# A. Occlusion Flows

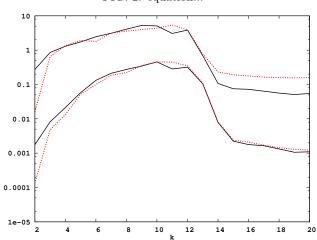
FIG. 1: equation...



$$\begin{split} \delta \boldsymbol{z}_{k-1} &\approx \boldsymbol{\Delta}_{\tau}^{\widetilde{\boldsymbol{z}}_{k-2} \cdot \boldsymbol{m}[\widetilde{\boldsymbol{x}}_{k-2}]} + \boldsymbol{z}_{k-2} \cdot \boldsymbol{m}' \left[ \widetilde{\boldsymbol{x}}_{k-2} \right] \cdot \delta \boldsymbol{x}_{k-2} \\ &+ \delta \boldsymbol{z}_{k-2} \cdot \boldsymbol{m} \left[ \widetilde{\boldsymbol{x}}_{k-2} \right] \quad (16) \end{split}$$

$$\|\delta \boldsymbol{z}_{k-1}\| \lesssim \|\boldsymbol{z}_{k-2}\| (\tau \|\mathbf{m} [\widetilde{\boldsymbol{x}}_{k-2}]\| + \|\delta \boldsymbol{x}_{k-2}\| \|\mathbf{m}' [\widetilde{\boldsymbol{x}}_{k-2}]\| ) \quad (17)$$

FIG. 2: equation...



$$\|\boldsymbol{z}_k\| \to \sqrt{\kappa(\boldsymbol{s})}$$
 (18)

C. Scaling

D. Comments

Pictures of the spamm structure

VII. CONCLUSION