

# Quantum Information and Computing 2022 - 2023

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#### matrix-matrix multiplication

#### **Theory**

matrix multiplication is a binary operation that produces a matrix from two matrices. For matrix multiplication, the number of columns in the first matrix must be equal to the number of rows in the second matrix.

#### **Code development**

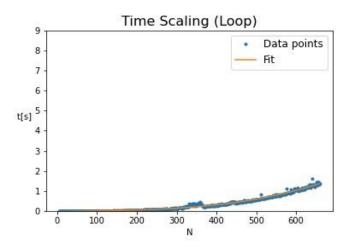
- We create a function perform matrix multiplication through a loop1, loop2 and matmul method.
- Store the matrix in csv file and run them in python code.

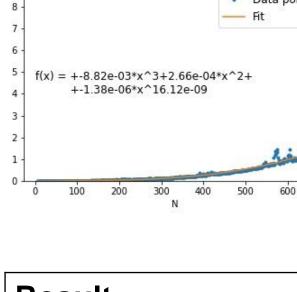
```
function matrix multiplication transposed (matrixA, matrixB) result(matr
     integer :: ii, jj, kk
     logical :: check
     real*4, dimension(:,:) :: matrixA, matrixB
     real*4, dimension(size(matrixA,2),size(matrixB,1)) :: matrixC
Check if multiplication is possible (shapes)
     if (size(matrixA,2) .eq. size(matrixB,1)) then
         check = .TRUE.
         print*, "Input matrices cannot be multiplied"
         check = .FALSE.
     end if
Matrix multiplication
     if (check .eqv. .TRUE.) then
         do kk = 1, size(matrixA,1), 1
             do jj = 1, size(matrixB,2), 1
                 do ii = 1, size(matrixB,1), 1
                     if (kk == 1) then
                         matrixC(ii,jj) = 0
                      end if
                     matrixC(ii,jj) = matrixC(ii,jj) + matrixA(ii,kk)*mat
          enddo
     end if
 end function matrix multiplication transposed
```



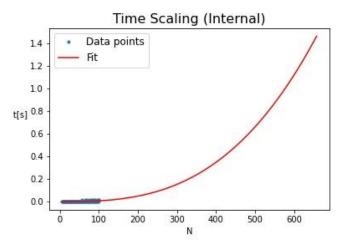
### matrix-matrix multiplication

Data points





Time Scaling (Loop2)



#### Result

- We can assume that the matrix multiplication operation scales line  $O(n^3)$ .



#### **Theory**

Random Hermitian matrices, drawn from one-cut regular unitary invariant ensembles, converge in law to Gaussian multiplicative chaos measures. We prove this in the so-called L2L2-phase of multiplicative chaos

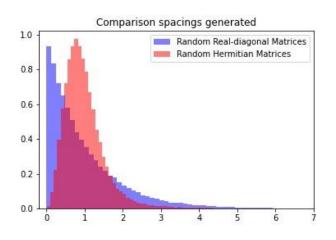
#### **Code Development**

- Study of the P(s) distribution, where  $s_i$  are the normalized spacings between eigenvalues:

$$egin{bmatrix} a_{11} & a_{12} & a_{13} \ a_{21} & a_{22} & a_{13} \ a_{31} & a_{32} & a_{33} \ \end{bmatrix}$$
 ZHEEV() (Lapack)  $\lambda_1,\lambda_2,\lambda_3$   $\lambda_1<\lambda_2<\lambda_3$   $\longrightarrow$   $s_i=\Delta\lambda_i/\overline{\Delta}\lambda_1$ 

- We can exploit the 'BLAS' and 'LAPACK' -libraries which contains 'ZHEEV'.
- In python, we will fit the function P(s) using curve fit from scipy.optimize

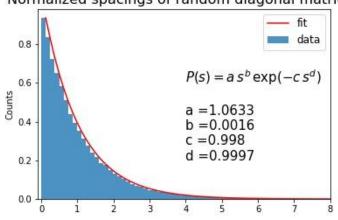
## Random matrix theory



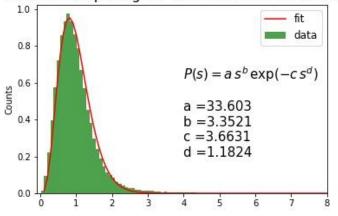
#### Result

- We perform the fit of normalize spacing of random diagonal and Hermitian matrices.

#### Normalized spacings of random diagonal matrices



#### Normalized spacings of random hermitian matrices







## Università degli Studi di Padova

# Thanks for the attention