

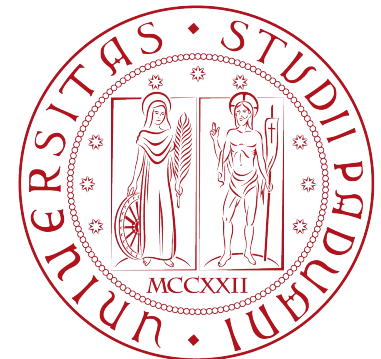
Presentation of Assignment 3

Quantum Information and Computing, A.Y. 2022/2023

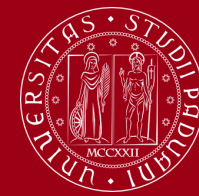
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Matrix multiplication scalings



```
EX3a_script.py x
1 import os
2 import math as m
3
4 # dimension grid parameters
5 N_min, N_max = 100, 12000
6 mult_factor = m.sqrt(2.0)
7
8 # number of repetitions (for statistics)
9 N_rep = 3
10
11 # list of dimensions to try
12 dim_list = [int(N_min * mult_factor**exp) for exp in range(m.floor(m.log(N_max/N_min, mult_factor))+1)]
13
14 # list of different methods
15 method_list = ['naive', 'opt', 'builtin']
16
17 # run for different times the desired program (a.out)
18 for dim_ in dim_list:
19     for method_ in method_list:
20         for rep_ in range(N_rep):
21             os.system(f'./a.out {dim_} {dim_} {dim_} {dim_} {method_}')
```

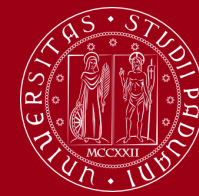
Execution
cycle

```
EX3a_Zinesi_CODE.f90 x
150
151 IF (TRIM(matmul_method) .EQ. "naive") THEN
152     ! naive function testing
153     CALL CPU_TIME(ti)
154     CALL my_MatMul_naive(AM,BM,CM)
155     CALL CPU_TIME(tf)
156
157
158     ! print performances string
159     WRITE (str, "(I5,'; ',I5,'; ',I5,'; ',I5,'; ',ES15.8E2)") &
160         SIZE(AM,1), SIZE(AM,2), SIZE(BM,1), SIZE(BM,2), tf-ti
161
162     ! write data into a dedicated file
163     OPEN(unit=10, file='results/'//TRIM(matmul_method)//'_performances.dat', access='APPEND')
164     WRITE(10, '(A)') TRIM(str)
165     CLOSE(unit=10)
```

The Python script “EX3a_script.py” executes for multiple times the code compiled from “EX3a_Zinesi_CODE.f90”. The result of each execution is appended to an output file based on the matrix multiplication method. This approach allows to freely choose the number of repetitions in the Python script without recompiling the source code.

```
paolozinesi@MBP-di-Paolo Assignment3/EX3a » ./a.out 200 200 200 200 opt
paolozinesi@MBP-di-Paolo Assignment3/EX3a » ./a.out 300 300 300 300 builtin
```

Matrix multiplication scalings

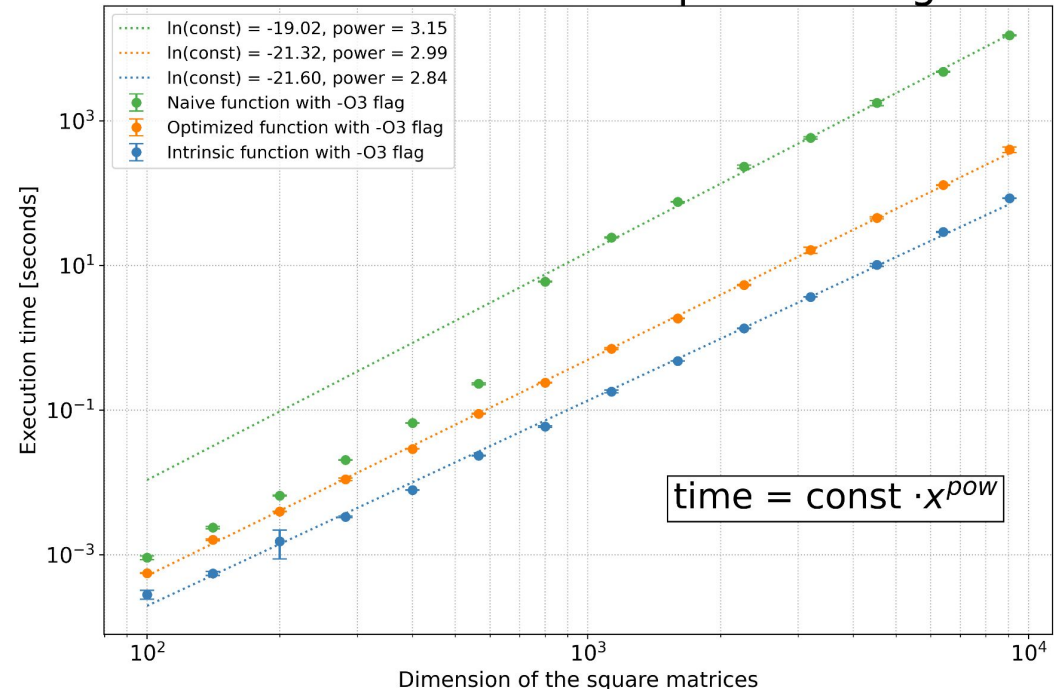


The Python script
“EX3a_plots.py” plots and
fits the execution times
obtained previously.

Only sizes ≥ 800 for the
naive method have been
considered in the linear fit.

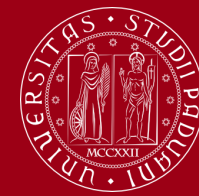
The optimization
performed by the compiler
causes the observed
difference in the naive
method scalings for small
and large matrices, even if
the two regimes only differ
by a multiplicative
constant.

Performances of matrix multiplication algorithms



```
EX3a_plots.py X
57 # fit
58 if type_=='naive':
59     # restrict domain
60     popt, pcov = curve_fit(f=pow_func, xdata=np.log(plot_df.loc[plot_df['size']>=800, 'size']),
61                           ydata=np.log(plot_df.loc[plot_df['size']>=800, 'time_mean']))
62 else:
63     popt, pcov = curve_fit(f=pow_func, xdata=np.log(plot_df['size']), ydata=np.log(plot_df['time_mean']))
64 ln_const, power = popt
65
66 # plot
67 x0 = np.linspace(min(plot_df['size']), max(plot_df['size']), 1000).reshape(-1,1)
68 ax.plot(x0, np.exp(pow_func(np.log(x0), ln_const=ln_const, pow=power)), ls='dotted', c=col_dict[type_],
69        label=f"ln(const) = {ln_const:.2f}, power = {power:.2f}")
```

Improvements to DCmatrix



```
≡ DCmatrix_mod.f90 ×
245      ! This subroutine stores into the variable 'M_rand' a random DCmatrix following a given input distribution.
246      ! The LAPACK library ZLATMR is employed to generate such random matrix.
247      ! Only double complex square matrices can be generated with this function.
248      !
249      ! inputs:
250      ! - M_rand [DCmatrix]: DCmatrix to fill with random values
251      !
252      !
253      !
254      !
255      !
256      !
257      !
258      !
259      !
260      !
261      !
262      !
263      !
264      !
265      !
266      !
267      !
268      SUBROUTINE RandMat(M_rand, dist, iseed, sym)
269
270      ! filling of DCmatrix using LAPACK routine ZLATMR
271      CALL ZLATMR(size, size, dist, iseed, sym, diag, 6, 1.D0, 1.D0, 'F', &
272      'N', DL, 0, 1.D0, DR, 0, 1.D0, 'N', ipivot, size, size, 0.D0, -1.0D0, 'N', &
273      M_rand%elem, size, iwork, info)
274
275      SUBROUTINE eigenvaluesMat(M, eigenvals)
276
277      ! call eigenvalues subroutine
278      CALL ZHEEV('N', 'U', size, M%elem, size, eigenvals, work, lwork, rwork, info)
```

Two subroutines has been added to DCmatrix module to wrap and simplify calls to **ZLATMR** and **ZHEEV** LAPACK subroutines. In particular, the eigenvalue subroutine for Hermitian matrices might be reused in future code.

“EX3b_Zinesi_CODE.f90” tests these new functionalities.

```
paolozinesi@MBP-di-Paolo Assignment3/EX3b » ./a.out 10
Normalized spacings printed on 'results/norm_spac.dat'
```

Eigenvalues spacings



$\{\lambda_i\}$: Eigenvalues of a random matrix in increasing order

$\Lambda_i = \lambda_{i+1} - \lambda_i$: Eigenvalue spacings

$s_i = \frac{\Lambda_i}{\langle \Lambda_i \rangle}$: Normalized eigenvalue spacings

“EX3c_Zinesi_CODE.f90” uses the DCmatrix routines “**RandMat**” and “**eigenvaluesMat**” to compute the normalized spacings distribution for random Hermitian matrices. The normalized spacings distribution for random real diagonal matrices are computed, instead, by using directly **DLARN** and **DLASRT** LAPACK subroutines. Results are appended to a dedicated output file.

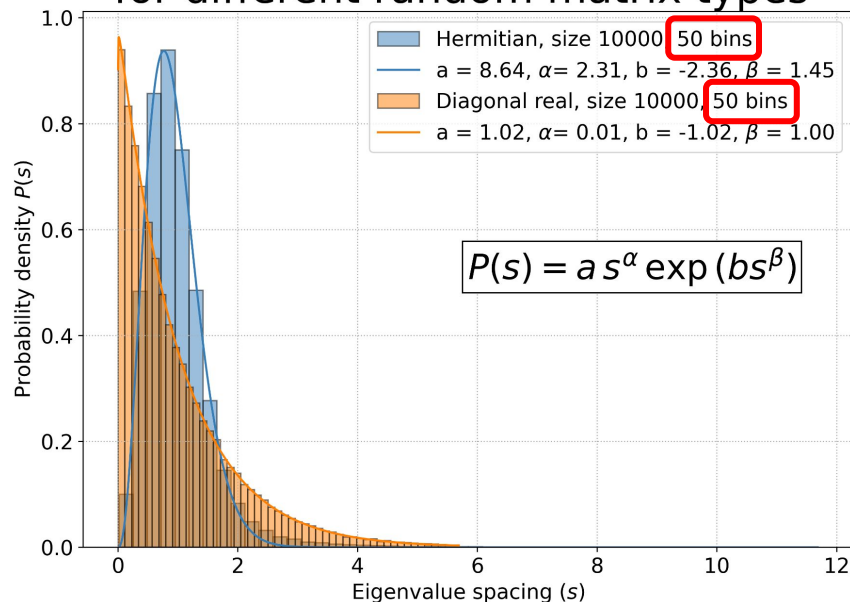
The real and imaginary parts of the complex Hermitian matrix entries are chosen to be **uniform in [-1,1]**. Similarly, the elements of the diagonal real random matrix are uniformly distributed in [-1,1]. Since the mean value of these random entries is zero, there is no need to neglect the largest eigenvalue.

```
paolozinesi@MBP-di-Paolo Assignment3/EX3c » ./a.out 2000 HS → Hermitian
paolozinesi@MBP-di-Paolo Assignment3/EX3c » ./a.out 2000 DS → Diagonal real
```

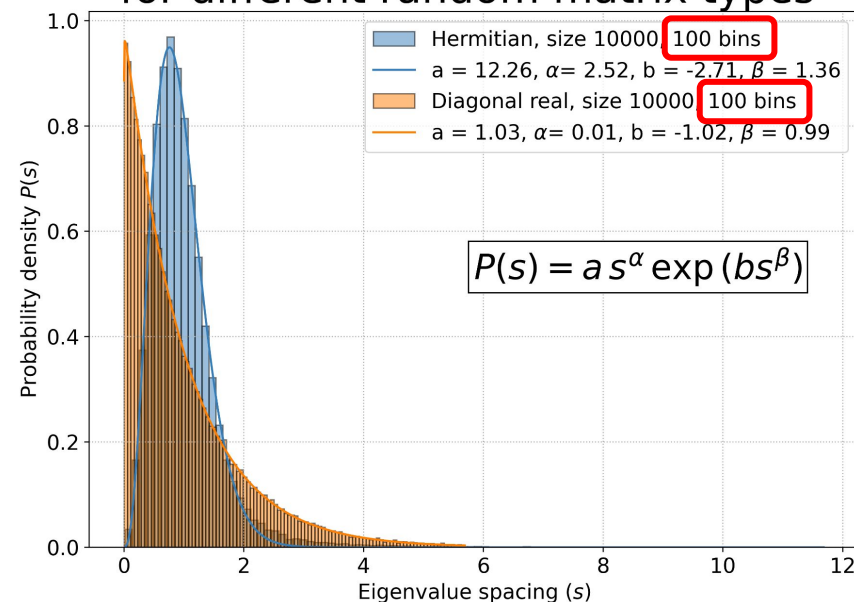

Eigenvalues spacings



Distribution of normalized spacings for different random matrix types



Distribution of normalized spacings for different random matrix types



The normalized spacings distribution is plotted and fitted with the Python script “EX3c_plots.py”. Different bin choices lead to different optimal fit parameters in the Hermitian case.

100 bins	a	α	b	β
Hermitian	12 ± 2	2.52 ± 0.09	-2.71 ± 0.17	1.36 ± 0.05
Real diag.	1.03 ± 0.02	0.013 ± 0.006	-1.02 ± 0.02	0.99 ± 0.01