Assignment 3

https://github.com/jchryssanthacopoulos/quantum_information/tree/main/assignment_3

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Exercise 1: Matrix Multiplication Scaling



- Matrix multiplication program extended to parse command-line arguments using get_command_argument
- Python script written in Jupyter notebook to make subprocess calls to Fortran program to get runtimes

```
$ compiled/exercise 1 00 \
--mat mul method matmul --num rows 3 \
--num_cols 3 --num_inner_dim 4 --debug
mat mul method = matmul
num_rows = 3
num cols = 3
num inner dim = 4
Running in debug mode ...
Matrix A =
  0.43 0.39
               0.04
  0.42 0.95 0.88
                     0.55
  0.01 0.19 0.47
                     0.62
Matrix B =
  0.10 0.37 0.73
  0.96 0.39 0.38
  0.03 0.53 0.23
  0.21 0.36 0.69
Product =
        0.39 0.59
  1.10 1.19 1.25
         0.55 0.62
Elapsed time = 7.0000000000E-06
```

```
lef get_run_time(mat_mul_method, flag, mat_dim):
    """Get the run time for the given matrix multiplication method,
    optimization flag, and matrix dimension.
    """
    run_params = [
        f"(program_base_name)_{flag}",
        ""--mat_mul_method", mat_mul_method,
        "--num_rows", str(mat_dim),
        "--num_cols", str(mat_dim),
        "--num_inner_dim", str(mat_dim)
]

output = subprocess.run(
    run_params, stdout=subprocess.PIPE, encoding='ascii'
)
lines = output.stdout.split('\n')
return float(lines[4].split('\n')
```

Exercise 1: Matrix Multiplication Scaling



Linear fit of log run time to log matrix size performed with numpy's polyfit

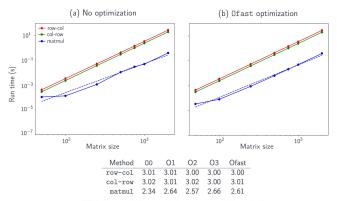


Table: Polynomial time complexity, given by coefficient a in fit $\log T = a \log N + b$

Exercise 2: Eigenproblem



- Hermitian matrix created with random number generator
- Eigenvalues computed using zheev

```
function rand hermitian matrix(n) result(H)
   integer n
   integer ii. ii
   complex*16 H(n, n)
   do ii = 1, n
       do 11 = 1, 11
           if (ii /= jj) then
               ! sample random complex numbers off the diagonal
               H(ii, jj) = cmplx(rand(8)*2 - 1, rand(8)*2 - 1)
               H(jj, ii) = conjg(H(ii, jj))
               ! sample real numbers on the diagonal
               H(ii, ii) = rand(0)*2 - 1
           end if
       end do
   end do
end function
```

```
I compute eigenvalues in ascending order
call lahew('N', 'U', ndin, H, ndin, eigvals, work, twork, rwork, info)
! compute eigenvalue spacings and average
ave_delta_eigvals = (eigvals(ndin) - eigvals(ii) / (ndin - 1)
do ii = 1, ndin - 1
norm_eigval_spacings(ii) = (eigvals(ii + 1) - eigvals(ii)) / ave_delta_eigvals
end do
```

Exercise 3: Random Matrix Theory



- Distribution of normalized spacings P(s) produced for Hermitian and diagonal matrix
- Fit to $P(s) = as^{\alpha} \exp(bs^{\beta})$ performed using scipy's curve_fit function

```
def func(x, a, b, alpha, beta):
    return a * x ** alpha * np.exp(b * x ** beta)
popt, pcov = curve_fit(func, bin_centers, norm_count)
popt
array([13.79272512, -2.82466922, 2.59653888, 1.32155578])
```

Matrix	а	b	α	β
hermitian	13.792 ± 2.266	-2.825 ± 0.166	2.597 ± 0.091	1.322 ± 0.043
diagonal	1.017 ± 0.007	-1.018 ± 0.008	0.005 ± 0.002	0.987 ± 0.005

Exercise 3: Random Matrix Theory



- Histograms produced with matrix size N = 1000 and number of bins $n_{\text{bins}} = 100$
- Results agree with random matrix theory (cf. Edelman and Rao, *Random Matrix Theory*, *Acta Numerica*, 2005, pp. 41)

