# Rust Quantum Library

The quantum library in Rust contains several useful measures of quantum entanglement that are calculated from the density matrix representing a quantum state. This library is generalized for any density matrix with dimension of  $2^n$ . The Rust library uses ndarray and the ndarray\_linalg packages.

We can run Rust from a Jupyter notebook by installing EvCxR Jupyter kernel. The Jupyter notebook runs with the **plotters** crate. Setup can be found from https://datacrayon.com/posts/programming/rust-notebooks/setup-anaconda-jupyter-and-rust. Miniconda is not necessary, Anaconda will work the same. Be careful when installing the actual kernel in the tutorial (shown below) - it is important to check on "other installation methods" in order for proper installation for Linux, Windows, and Mac.

### Install the EvCxR Jupyter Kernel

Now we'll install the EvCxR Jupyter Kernel. If you're wondering how it's pronounced, it's been mentioned to be "Evic-ser". This is what will allow us to execute Rust code in a Jupyter Notebook.

You can get other installation methods methods for EvCxR if you need then, but we will be using:

```
cargo install evcxr_jupyter --version 0.5.3 evcxr_jupyter --install
```

Below are the various quantum functions along with the functions that perform linear algebra operations.

## 1 Quantum functions

### 1.1 Create a density matrix $\rho$ (create\_density\_matrix)

**Input:** The wavefunction expressed as a column vector of coefficients

Input type: pub type VecC64 = ndarray::Array1 < c64 >

**Input Example:** For the Bell state  $\phi_+ = \frac{1}{\sqrt{2}}(|00\rangle + |11\rangle)$ ,

 $let \ norm\_const = 1./2\_f64.sqrt();$ 

 $let \ bell\_phi\_plus = array![c64::new(norm\_const \ , \ 0.0) \ , \ c64::new(0.0 \ , \ 0.0) \ , \ c64::new(0.0 \ , \ 0.0) \ , \ c64::new(0.0 \ , \ 0.0)];$ 

**Output:** The density matrix, which is square and Hermitian.

Output type: pub type MatrixC64 = ndarray::Array2<c64>

Output Example: dens matrix for phi\_+ =

[[0.4999999999999999999]i, 0+0i, 0+0i, 0.4999999999999999i].

[0+0i, 0+0i, 0+0i, 0+0i],

[0+0i, 0+0i, 0+0i, 0+0i],

The density matrix is a more general way of representing the state of a quantum system. It is represented as  $\rho = |\psi\rangle\langle\psi|$ , where the wavefunction  $|\psi\rangle$  is represented as a column vector.

With this same example, the function performs the operation

$$\rho = |\phi_{+}\rangle \langle \phi_{+}| = \frac{1}{2} \begin{pmatrix} 1 \\ 0 \\ 0 \\ 1 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 & 1 \end{pmatrix}^{*} = \frac{1}{2} \begin{pmatrix} 1 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 1 \end{pmatrix}$$

### 1.2 Find purity (find\_purity)

```
Input: ρ (Density Matrix)
Input Type: pub type MatrixC64 = ndarray::Array2<c64>
Input Example:
array![ [c64::new(0.5, 0.), c64::new(0., 0.)],
[c64::new(0., 0.), c64::new(0.5, 0.)] ]
```

Output: Float

Output Type: f64

Output Example: 0.5

Simply returns the trace of  $\rho^2$  as a f64 type. For an idea on where this purity lies within the range of possible values, print a statement such as println!("The purity lies between {} and 1", 1./(find\_dim(rho\_sqrd))). The find\_dim function finds the dimension of the square matrix. For example, a 4 x 4 matrix has dim = 4.

#### 1.3 Find fidelity (find\_fidelity)

```
Input: Two density matrices \rho_1 and \rho_2

Input Type: pub type MatrixC64 = ndarray::Array2<c64>
Input Example: let rho_1 = array![
[c64::new(0.25, 0.0), c64::new(0.0, 0.0), c64::new(0.0, 0.0), c64::new(0.0, 0.0)],
[c64::new(0.0, 0.0), c64::new(0.25, 0.0), c64::new(0.0, 0.0), c64::new(0.0, 0.0)],
[c64::new(0.0, 0.0), c64::new(0.0, 0.0), c64::new(0.25, 0.0), c64::new(0.0, 0.0)],
[c64::new(0.0, 0.0), c64::new(0.0, 0.0), c64::new(0.0, 0.0)],
[c64::new(0.0, 0.0), c64::new(0.0, 0.0), c64::new(0.0, 0.0)]]
[c64::new(0.0, 0.0), c64::new(0.0, 0.0)]]
```

Output: A number between 0 and 1

Output Type: f64

Output Example: 1

This function is a distance measurement of two quantum states  $\rho_1$  and  $\rho_2$ . It is expressed as  $F = tr\sqrt{\sqrt{\rho_1}\rho_2\sqrt{\rho_1}}$ . This function calls  $find\_sqr\_root\_of\_matrix$  to calculate  $\sqrt{\rho_1}$  and the overall square root of  $\sqrt{\rho_1}\rho_2\sqrt{\rho_1}$ .

## 1.4 Find concurrence (find\_concurrence)

**Input:** The density matrix  $\rho$ 

Input Type: pub type MatrixC64 = ndarray::Array2<c64>

**Input Example:** The state  $\cos(\theta) |00\rangle + \sin(\theta) |11\rangle$  for  $\theta = 30^{\circ}$  is represented by the vector:

pub const THETA:  $f64 = PI^*(30./180.)$ ;

let psi\_part\_entangled: VecC64 = array![c64::new(THETA.cos(), 0.0), c64::new(0.0, 0.0), c64::new(0.0, 0.0), c64::new(THETA.sin(), 0.0)];

Output: A number between 0 and 1

Output Type: f64

Output Example: 0.866 (which is the rounded from  $\sqrt{3}/2$ )

For two qubits, concurrence  $C(\rho) = max(0, \lambda_1 - \lambda_2 - \lambda_3 - \lambda_4)$ , where  $\lambda_1, ... \lambda_4$  are the eigenvalues in decreasing order (ie  $\lambda_1$  is the highest eigenvalue) of the matrix  $R = \sqrt{\sqrt{\rho}\tilde{\rho}\sqrt{\rho}}$ .

In this case,  $\tilde{\rho} = (\sigma_y \otimes \sigma_y) \rho^* (\sigma_y \otimes \sigma_y)$ .  $\rho^*$  is the complex conjugate of  $\rho$ , and  $\sigma_y$  is the Pauli-y spin matrix.

This function has been extended from 2 qubits (4 x 4 density matrix) to 3 or more qubits. The function will give wrong answers for odd numbers of qubits.

The extension uses  $C(\rho) = max(0, \lambda_1 - \sum \lambda_i)$ , where i goes from 2 to the density matrix dimension. The dimension of the density matrix is  $2^n$ , where n is the number of qubits. Also in this case  $\tilde{\rho} = (\sigma_y \otimes^n \sigma_y) \rho^*(\sigma_y \otimes^n \sigma_y)$ .

The tensor product operation for  $\otimes^n$  is called using the  $find\_tensor\_product$  function, and uses an iterator with the fold method to perform the tensor product n times.

## 1.5 Find trace norm (find\_trace\_norm)

**Input:** The density matrix  $\rho$ 

Input Type: pub type MatrixC64 = ndarray::Array2<c64>

Input Example:

For the Bell state  $\phi_+ = \frac{1}{\sqrt{2}}(|00\rangle + |11\rangle),$ 

let norm\_const = 1./2\_f64.sqrt();

let bell\_phi\_plus = array![c64::new(norm\_const , 0.0) , c64::new(0.0 , 0.0) , c64::new(0.0 , 0.0) , c64::new(norm\_const , 0.0)];

**Output:** A float from 0 to 1

Output Type: f64

Output Example: 1

The trace norm is expressed as  $||\rho||_1 = tr\sqrt{\rho^{\dagger}\rho}$ .  $\rho^{\dagger}$  is the complex conjugate and transpose of the matrix  $\rho$ . The function calls  $find\_sqr\_root\_of\_matrix(\rho^{\dagger}\rho)$ . For Hermitian, normalized density matrices, the output is always 1.

### 1.6 Find negativity (find\_negativity)

**Input:** The density matrix  $\rho$ 

Input Type: pub type MatrixC64 = ndarray::Array2<c64>

Input Example: For a 5 qubit maximally mixed density matrix,

let rho\_mixed\_diag:  $VecC64 = Array::from\_elem(32, c64::new(1./32., 0.0));$ 

let rho\_mixed: MatrixC64 = MatrixC64::from\_diag(rho\_mixed\_diag);

Output: A number between 0 and 1

Output Type: f64

Output Example: 0.0

Th negativity can be expressed as  $N = \frac{||\rho^{\Gamma_A}||_1-1}{2}$  where  $\rho^{\Gamma_A}$  is the partial transpose of a substate of  $\rho$ .  $||\rho^{\Gamma_A}||_1$  is the trace norm of that partial transpose. Although the trace norm for density matrices is 1, the trace norm of the partially transposed matrix can vary from 0 to 1. The Peres-Horodecki criterion separates  $\rho$  into the tensor product of two states A and B.  $\rho^{\Gamma_A}$  is called with the function  $find\_partial\_transpose(\rho)$ .

## 1.7 Find log negativity (find\_log\_negativity)

**Input:** The density matrix  $\rho$ 

**Input Type:** pub type MatrixC64 = ndarray::Array2 < c64 >

**Input Example:** For the Bell State  $\phi_{-} = \frac{1}{\sqrt{2}}(|00\rangle - |11\rangle)$ :

pub const NORM\_CONST: f64 = 1./std::f64::consts::SQRT\_2;

let bell\_phi\_minus\_vec:  $VecC64 = array![ c64::new(NORM\_CONST, 0.0) , c64::new(0.0, 0.0) ,$ 

 $c64::new(0.0, 0.0), c64::new(-NORM_CONST, 0.0)$ ;

let rho\_bell\_psi\_plus = create\_density\_matrix(BELL\_PSI\_PLUS\_VEC);

Output: A number between 0 and 1

Output Type: f64

Output Example: 0.0

This function lets  $N = find\_negativity(\rho)$ . Then it takes  $log_2(2N+1)$ .

## 1.8 Find Schmidt number (find\_schmidt\_number)

Input: The joint spectral intensity (JSI) matrix

**Input Type:** pub type MatrixF64 = ndarray::Array2<f64>

Input Example: A 200 x 200 JSI (located inside the examples folder)

Output: An entanglement measurement of the JSI.

Output Type: f64

Output Example: 6.74 (rounded)

This computation is done in several steps:

1) Take the square root of the elements of the JSI (joint spectral intensity) to find the JSA (joint spectral amplitude). The JSI is a matrix of floats from SPDCalc.

2) Find the singular value decomposition (SVD) of the JSA. A SVD calculation decomposes the into three matrices,  $USV^T$ .

3) A normalization constant is found by taking the inverse of the sum of the squares of the singular values of the S matrix (eigenvalues), i.e.  $A = 1/\sum \lambda_i^2$ .

4) The S matrix is renormalized by multiplying the  $\sqrt{A} * S$ .

5) From the normalized S matrix, take the sum of the eigenvalues to the fourth power, i.e.  $\sum \lambda_{i,norm}^4$ . The Schmidt number is  $k = 1/\sum \lambda_{i,norm}^4$ .

## 1.9 Create two source Hong-Ou-Mandel graph (find\_two\_source\_hom)

**Inputs:** A signal vector of frequency modes  $\omega$ , an idler vector of modes  $\omega$ , a joint spectral amplitude (JSA) of modes  $\omega$ , and a vector of time intervals dt in femtoseconds.

Input Types: signal: VecF64, idler: VecF64, jsa: MatrixC64, dt: VecF64

Output: A graph of time intervals (dt) on the x axis and coincidence probabilities on the y axis.

Output Example An example is located in the "two\_source\_hom\_plot.ipynb" under qm/examples.

When calculating the two-source Hong-Ou-Mandel graph, we consider four photons from two identical crystals, i.e. two signal photons and two idler photons. The relevant integral to compute is

$$\int_{\omega_s} \int_{\omega_i} \int_{\omega'_s} \psi(\omega_s, \omega_i, \omega'_s, \omega'_i) - e^{i\phi} \psi(\omega'_s, \omega_i, \omega_s, \omega'_i) \ d\omega'_i \ d\omega'_s \ d\omega_i \ d\omega_s.$$

The wave function  $\psi$  is equivalent to the joint spectral amplitude (JSA) for all four photons. Since the photons come from two separate sources, this joint amplitude is separable into two parts, so that

$$\psi(\omega_{s}, \omega_{i}, \omega_{s}', \omega_{i}') = JSA_{1}(\omega_{s}, \omega_{i}) \otimes JSA_{2}(\omega_{s}', \omega_{i}')$$

. Since the two sources are identical, we conclude that  $JSA_1 = JSA_2$ . So we only need to compute the JSA once and then simply reference different indices for the different signal and idler photons. Using this notation, the integrand becomes

$$JSA(\omega_s, \omega_i) \otimes JSA(\omega_s^{'}, \omega_i^{'}) - e^{i\phi}JSA(\omega_s^{'}, \omega_i) \otimes JSA(\omega_s, \omega_i^{'})$$

We can simply iterate over many sums rather than explicitly calculating this integral. The variable phi depends on time and wavelength difference between two photons of interest. For signal-signal interference,  $\phi_{ss} = 2\pi\Delta t(\frac{1}{\lambda_s} - \frac{1}{\lambda_i})$ . For idler-idler interference,  $\phi_{ii} = 2\pi\Delta t(\frac{1}{\lambda_s'} - \frac{1}{\lambda_i'})$ . For signal-idler interference,  $\phi_{si} = 2\pi\Delta t(\frac{1}{\lambda_s} - \frac{1}{\lambda_i'})$ .

In the code, we let A, B, C, D stand for the different JSA configurations, between  $\omega_s$  and  $\omega_i$ ,  $\omega_s'$  and  $\omega_i'$ ,  $\omega_s'$  and  $\omega_i$ , and  $\omega_s$  and  $\omega_i'$  respectively.

The find\_two\_source\_hom\_norm function normalizes the coincidence probabilities.

#### 1.10 Find two source Hong-Ou-Mandel norm (find\_two\_source\_hom\_norm)

**Inputs:** A signal vector of frequency modes  $\omega$ , an idler vector of modes  $\omega$ , and a joint spectral amplitude (JSA) of modes  $\omega$ .

Input Types: signal: VecF64, idler: VecF64, jsa: MatrixC64

Output: A value that normalizes the two\_source\_hom for each time interval dt.

Output Type: f64

Output Example: 6.41e+61\_f64 (rounded)

Normalizes the coincidence probabilities in the two source HOM calculation.

## 2 Matrix Operations

## 2.1 Find the symmetric square root of a matrix (find\_symmetric\_square\_root)

```
Input: Any square, Hermitian matrix
```

**Input Type:** pub type MatrixC64 = ndarray::Array2<c64>

#### Input Example:

```
array![ [c64::new(0.0, 0.0), c64::new(0.0, 0.0), c64::new(0.0, 0.0)], [c64::new(0.0, 0.0), c64::new(0.0, 0.0), c64::new(0.0, 0.0)], [c64::new(0.0, 0.0), c64::new(0.0, 0.0)], [c64::new(0.0, 0.0)], [
```

Output: The square root of the matrix

Output Type: pub type MatrixC64 = ndarray::Array2<c64>

#### Output Example:

```
array! [ [c64::new(0.0, 0.0), c64::new(0.0, 0.0), c64::new(0.0, 0.0)], [ [c64::new(0.0, 0.0), c64::new(0.5, 0.0), c64::new(0.0, 0.5)], [ [c64::new(0.0, 0.0), c64::new(0.0, -0.5), c64::new(0.5, 0.0)], ];
```

Computes the symmetric square root of a Hermitian matrix. In the example above, the square

root of the matrix is equivalent to the matrix itself.

\*\*\*Bug in LAPACK involving numerical precision

For some of the entanglement measurements, the square root of the density matrix is required.

Using a LAPACK function such as matrix.ssqrt(UPLO:Lower).unwrap() is the easiest and fastest

route, but LAPACK has a bug and fails to compute. It's suspected that the matrix needs to be

semi-positive definite (eigenvalues need to be  $\geq 0$ ) in order for it to work.

So, the function was written by decomposing the Hermitian matrix as  $M = SDS^{-1}$ , where

S and  $S^{-1}$  are complex unitary matrices and D is a real diagonal matrix. S is a matrix of the

eigenvectors of M, and D is has the eigenvalues of M in the diagonal. Since D is diagonal, one can

take the square of the elements in order to find  $\sqrt{D}$ . So,  $\sqrt{M} = S\sqrt{D}S^{-1}$ .

But, one problem is that again, when doing this decomposition, the matrix M needs to be

semi-positive definite. It usually is, but LAPACK will sometimes find that one eigenvalue is a very

small negative number, such as  $2.8 \times 10^{-15}$ . A rescaling function is called in order to fix this issue.

The rescaling function takes matrix M as an input and finds its eigenvalues and eigenvectors,

and sets any negative eigenvalues to 0. Then, matrix M can be decomposed into  $M = SDS^{-1}$ .

The rescaling function returns (D, S) as a tuple. Then, the square root function returns  $\sqrt{M}$ 

 $S\sqrt{D}S^{-1}.$ 

Unfortunately after rescaling the function and calling LAPACK with matrix.ssqrt(UPLO:Lower).unwrap()

still fails in the computation. This is why (D, S) is returned as a tuple rather than a rescaled density

matrix.

2.2Find matrix dimension  $(find_-dim)$ 

**Input:** Square Matrix

**Input Type:** pub type MatrixC64 = ndarray::Array2<c64>

Input Example:

10

```
 \begin{split} & \text{array!} [ \text{ [c64::new(3., 1.) }, \text{ c64::new(-1., 1.) }], \\ & [ \text{c64::new(2., -1.) }, \text{ c64::new(-2., -1.)] }] \end{split}
```

Output: Integer

Output Type: i32

Output Example: 2

A simple function that calculates the length of a square matrix. For example, if a matrix is 8 x 8, this will return 8 as a type i32 (integer 32 bit).

### 2.3 Find partial transpose (find\_partial\_transpose)

**Input:** Any square matrix

**Input Type:** pub type MatrixC64 = ndarray::Array2<c64>

**Input Example:** let matrix: MatrixC64 = array!

```
 \begin{array}{l} [{\rm c64::new}(1.0\;,\,0.0)\;,\,{\rm c64::new}(2.0\;,\,0.0)\;,\,{\rm c64::new}(3.0\;,\,0.0)\;,\,{\rm c64::new}(4.0\;,\,0.0)\;]\;,\\ [{\rm c64::new}(5.0\;,\,0.0)\;,\,{\rm c64::new}(6.0\;,\,0.0)\;,\,{\rm c64::new}(7.0\;,\,0.0)\;,\,{\rm c64::new}(8.0\;,\,0.0)\;]\;,\\ [{\rm c64::new}(9.0\;,\,0.0)\;,\,{\rm c64::new}(13.0\;,\,0.0)\;,\,{\rm c64::new}(11.0\;,\,0.0)\;,\,{\rm c64::new}(12.0\;,\,0.0)]\;,\\ [{\rm c64::new}(10.0\;,\,0.0)\;,\,{\rm c64::new}(14.0\;,\,0.0)\;,\,{\rm c64::new}(15.0\;,\,0.0)\;,\,{\rm c64::new}(16.0\;,\,0.0)]\;]; \end{array}
```

Output: A square matrix

Output Type: pub type MatrixC64 = ndarray::Array2<c64>

Output Example:

```
 \left[ \text{c64::new}(1.0 \text{ , } 0.0) \text{ , } \text{c64::new}(2.0 \text{ , } 0.0) \text{ , } \text{c64::new}(3.0 \text{ , } 0.0) \text{ , } \text{c64::new}(7.0 \text{ , } 0.0) \right] , \\ \left[ \text{c64::new}(5.0 \text{ , } 0.0) \text{ , } \text{c64::new}(6.0 \text{ , } 0.0) \text{ , } \text{c64::new}(4.0 \text{ , } 0.0) \text{ , } \text{c64::new}(8.0 \text{ , } 0.0) \right] , \\ \left[ \text{c64::new}(9.0 \text{ , } 0.0) \text{ , } \text{c64::new}(10.0 \text{ , } 0.0) \text{ , } \text{c64::new}(11.0 \text{ , } 0.0) \text{ , } \text{c64::new}(12.0 \text{ , } 0.0) \right] , \\ \left[ \text{c64::new}(9.0 \text{ , } 0.0) \text{ , } \text{c64::new}(10.0 \text{ , } 0.0) \text{ , } \text{c64::new}(11.0 \text{ , } 0.0) \text{ , } \text{c64::new}(12.0 \text{ , } 0.0) \right] , \\ \left[ \text{c64::new}(9.0 \text{ , } 0.0) \text{ , } \text{c64::new}(10.0 \text{ , } 0.0) \text{ , } \text{c64::new}(11.0 \text{ , } 0.0) \text{ , } \text{c64::new}(12.0 \text{ , } 0.0) \right] , \\ \left[ \text{c64::new}(9.0 \text{ , } 0.0) \text{ , } \text{c64::new}(10.0 \text{ , } 0.0) \text{ , } \text{c64::new}(11.0 \text{ , } 0.0) \text{ , } \text{c64::new}(12.0 \text{ , } 0.0) \right] \right] , \\ \left[ \text{c64::new}(9.0 \text{ , } 0.0) \text{ , } \text{c64::new}(10.0 \text{ , } 0.0) \text{ , } \text{c64::new}(11.0 \text{ , } 0.0) \text{ , } \text{c64::new}(12.0 \text{ , } 0.0) \right] \right] , \\ \left[ \text{c64::new}(9.0 \text{ , } 0.0) \text{ , } \text{c64::new}(10.0 \text{ , } 0.0) \text{ , } \text{c64::new}(11.0 \text{ , } 0.0) \text{ , } \text{c64::new}(12.0 \text{ , } 0.0) \right] \right] , \\ \left[ \text{c64::new}(9.0 \text{ , } 0.0) \text{ , } \text{c64::new}(10.0 \text{ , } 0.0) \text{ , } \text{c64::new}(11.0 \text{ , } 0.0) \text{ , } \text{c64::new}(12.0 \text{ , } 0.0) \right] \right]
```

```
[c64::new(13.0, 0.0), c64::new(14.0, 0.0), c64::new(15.0, 0.0), c64::new(16.0, 0.0)]];
```

The partial transpose operation splits an even-numbered dimension matrix into four symmetrical blocks. The top right block and bottom left block are transposed, and then the matrix is put back together again.

### 2.4 Find tensor product (find\_tensor\_product)

```
Input: One square matrix with dim m \times m and another with dim n \times n
```

Input Type: pub type MatrixC64 = ndarray::Array2<c64>

**Input Example:** let matrix: MatrixC64 = array!

```
 [ c64::new(1.0 \ , \ 0.0) \ , \ c64::new(1.0 \ , \ 0.0) ] \ , \\ [ c64::new(1.0 \ , \ 0.0) \ , \ c64::new(-1.0 \ , \ 0.0) ] \ , \\ ] ;
```

**Output:** A square matrix with dim  $mn \times mn$ 

Output Type: pub type MatrixC64 = ndarray::Array2<c64>

Output Example:

```
 \begin{array}{l} [{\rm c64::new}(1.0\;,\,0.0)\;,\,{\rm c64::new}(1.0\;,\,0.0)\;,\,{\rm c64::new}(1.0\;,\,0.0)\;,\,{\rm c64::new}(1.0\;,\,0.0)\;]\;,\\ [{\rm c64::new}(1.0\;,\,0.0)\;,\,{\rm c64::new}(-1.0\;,\,0.0)\;,\,{\rm c64::new}(-1.0\;,\,0.0)\;,\,{\rm c64::new}(-1.0\;,\,0.0)\;]\;,\\ [{\rm c64::new}(1.0\;,\,0.0)\;,\,{\rm c64::new}(1.0\;,\,0.0)\;,\,{\rm c64::new}(-1.0\;,\,0.0)\;,\,{\rm c64::new}(-1.0\;,\,0.0)\;]\;,\\ [{\rm c64::new}(1.0\;,\,0.0)\;,\,{\rm c64::new}(-1.0\;,\,0.0)\;,\,{\rm c64::new}(-1.0\;,\,0.0)\;,\,{\rm c64::new}(-1.0\;,\,0.0)\;]\;]; \end{array}
```

Computes the tensor product of any matrix.