

Computational Quantum Mechanics

Prof. Gerson J. Ferreira

INFIS/UFU 2020/1

Eigenstates of 2D systems



2D systems

Now we want to consider a 2D Hamiltonian and solve: $H\psi(x, y) = E\psi(x, y)$

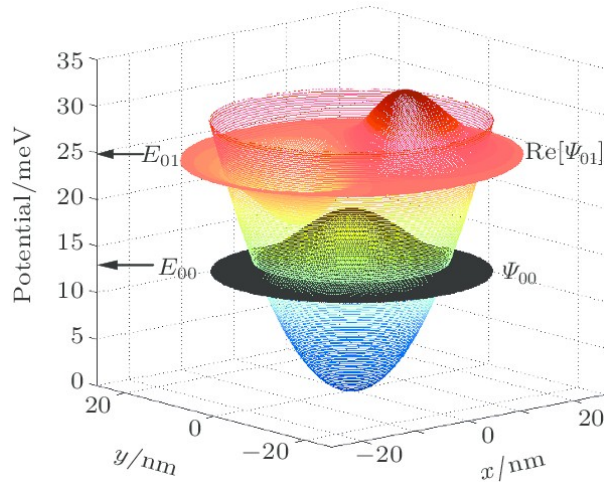
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[Edvinsson, R. Soc. Open Sci. 5, 180387 (2018)]



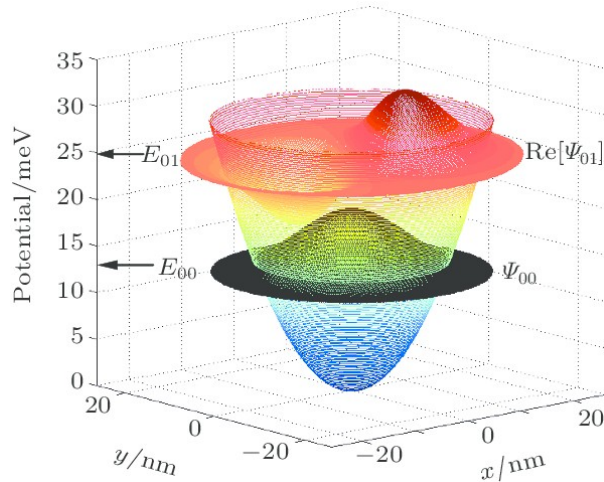
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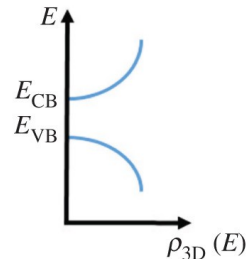
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[Song et al. Chinese Phys. B 21, 057302 (2012)]



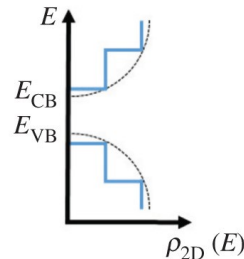
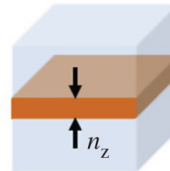
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no confinement
bulk
(3D materials)



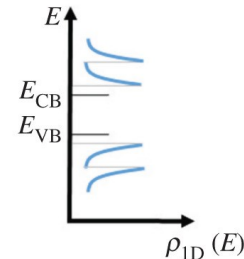
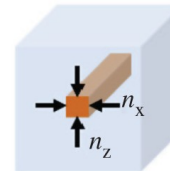
bulk

1D confinement
Q-well/ultrathin film
(2D materials)



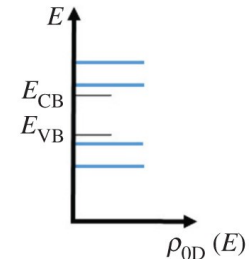
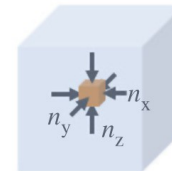
quantum well

2D confinement
Q-wire
(1D materials)



quantum wire

3D confinement
Q-dot
(0D material)



quantum dot



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Matrix representation for finite differences

$$H\psi(x, y) = E\psi(x, y) \quad H = -\frac{\hbar^2}{2m} \left(\frac{\partial^2}{\partial x^2} + \frac{\partial^2}{\partial y^2} \right) + V(x, y)$$



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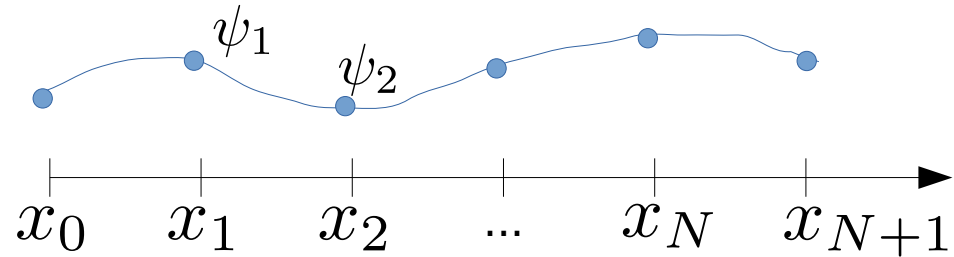


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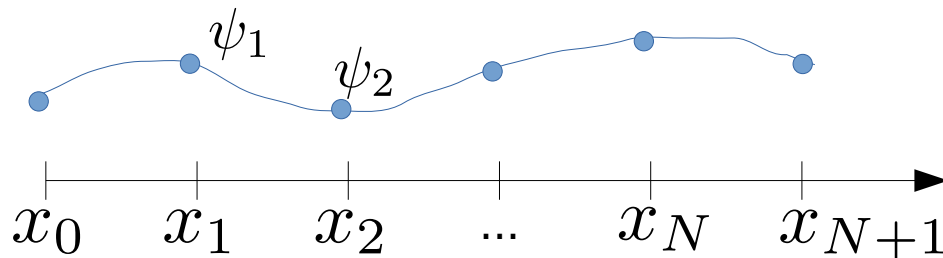
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$$\psi(x) \rightarrow \vec{\psi} = \begin{pmatrix} \psi_0 \\ \psi_1 \\ \psi_2 \\ \dots \\ \psi_{N-1} \end{pmatrix}$$



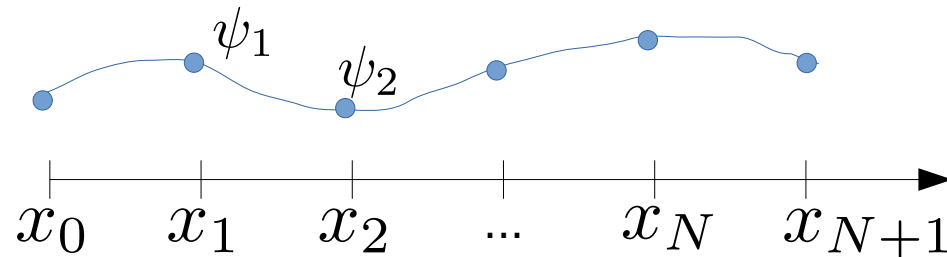
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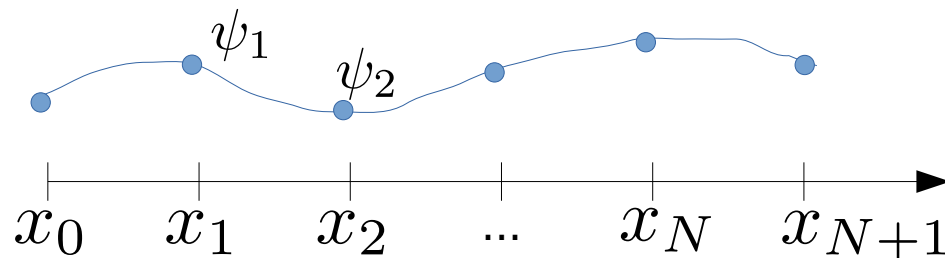
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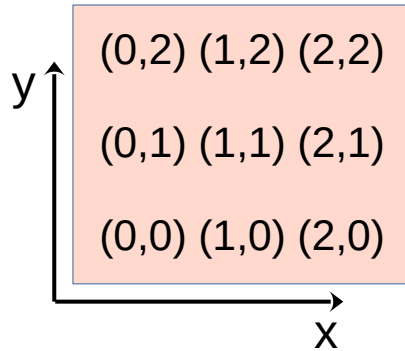
But now we have $\psi(x, y)$: how do we organize it as a vector?



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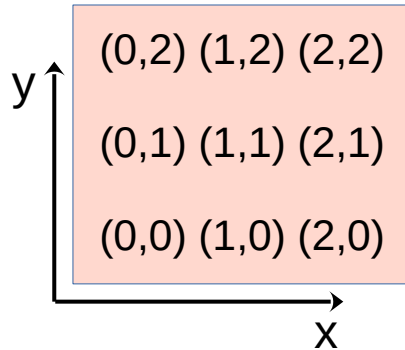


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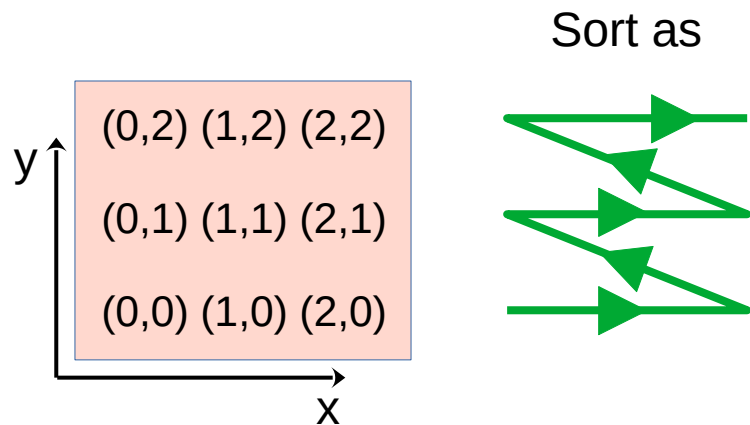


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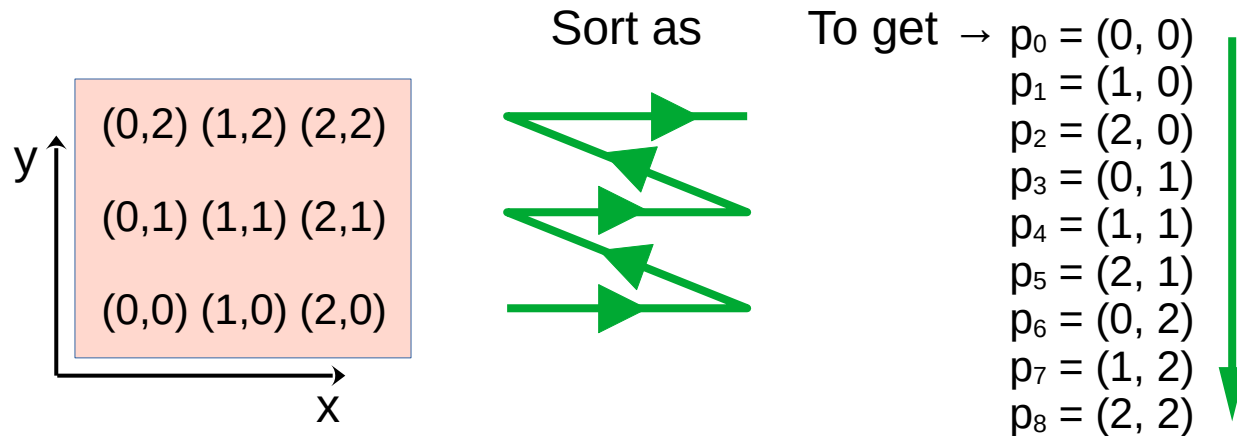


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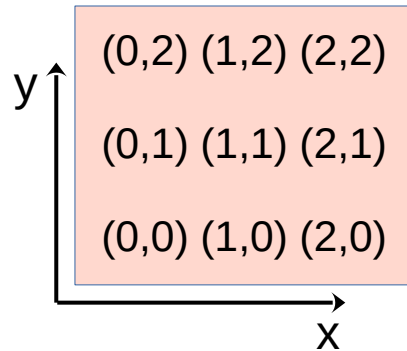


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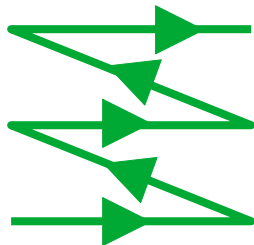
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Sort as



To get \rightarrow

$p_0 = (0, 0)$
 $p_1 = (1, 0)$
 $p_2 = (2, 0)$
 $p_3 = (0, 1)$
 $p_4 = (1, 1)$
 $p_5 = (2, 1)$
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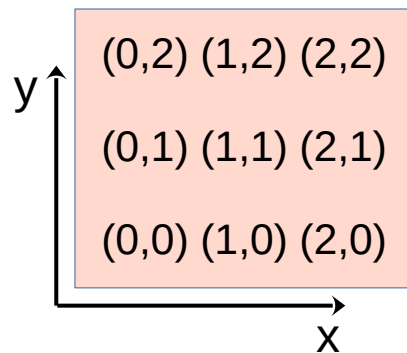


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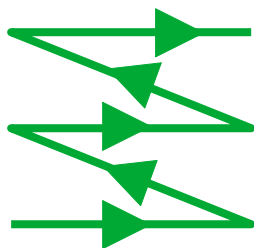
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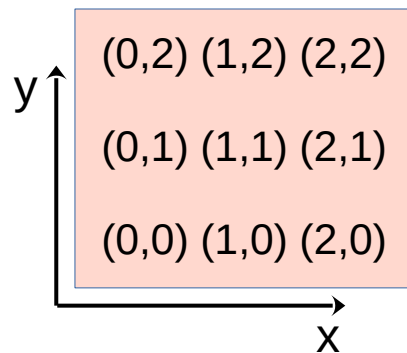


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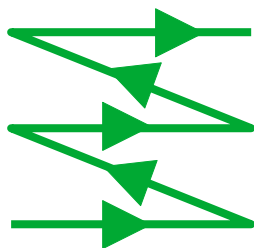
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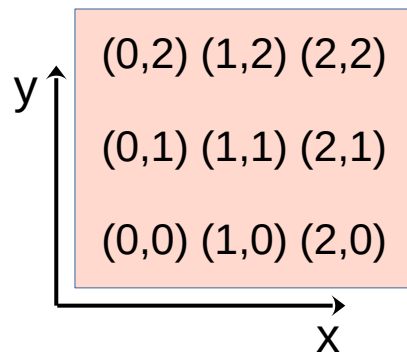


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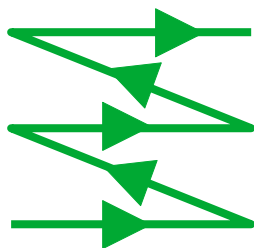
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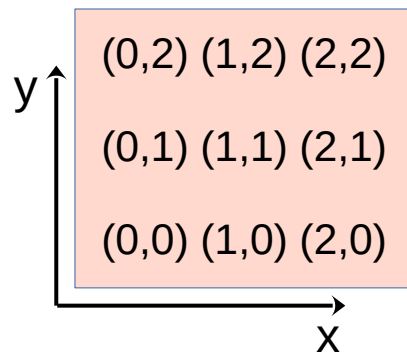


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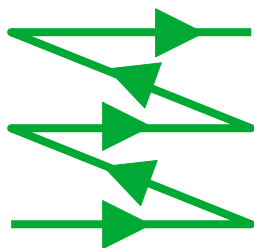
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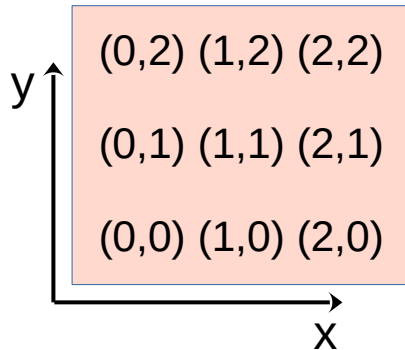
`np.kron(...)`



How these python commands work? **meshgrid** and **flatten**

meshgrid: convert vectors x, y into coordinate matrices

flatten: collapses the array into one dimension

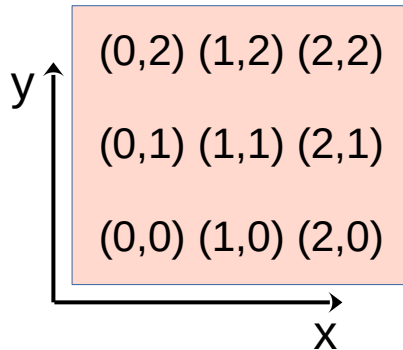


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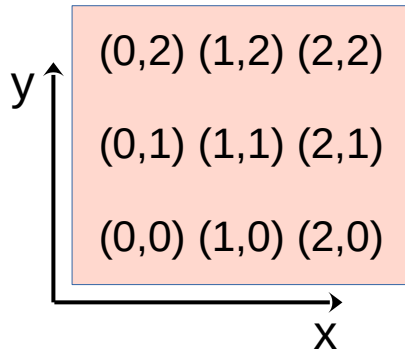
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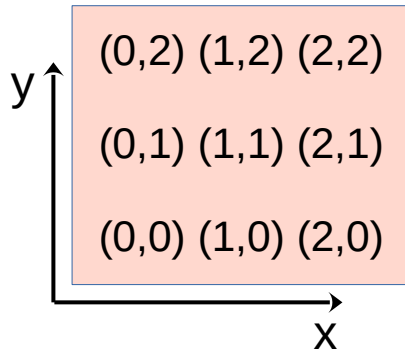
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Matrix:

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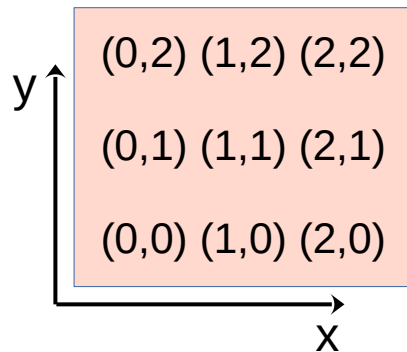
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Matrix:

	0,0,0
Matrix:	1,1,1
x → 0,1,2	2,2,2
x → 0,1,2	↑ ↑ ↑
x → 0,1,2	> > >



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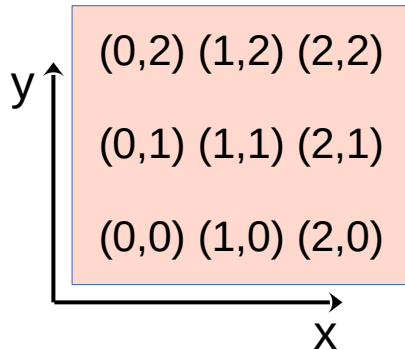
Matrix:

	0,0,0
x →	0,1,2
x →	0,1,2
x →	0,1,2

Matrix:

	1,1,1
	2,2,2
	>>>

The matrix elements form the pairs $p_n = (x_n, y_n)$



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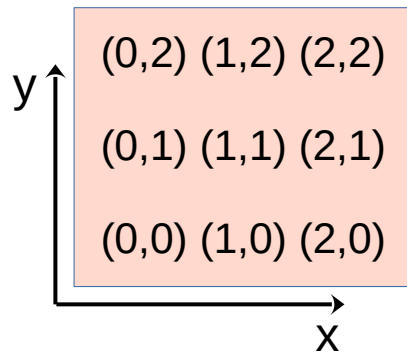
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Matrix:

	0	1	2
0	0,0,0	0,1,2	0,1,2
1	1,1,1	0,1,2	0,1,2
2	2,2,2	0,1,2	0,1,2

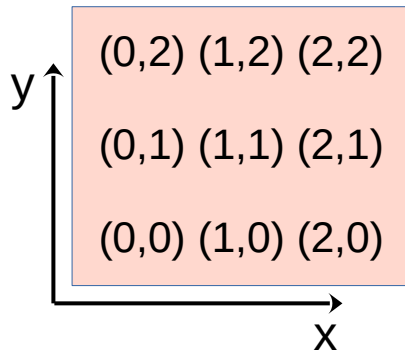
Matrix:

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0	0,0,0	0,1,2	0,1,2
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2	2,2,2	0,1,2	0,1,2

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Y = Y.flatten()
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Matrix:

	0	1	2
0	0	1	2
1	1	2	3
2	2	3	4

Matrix:

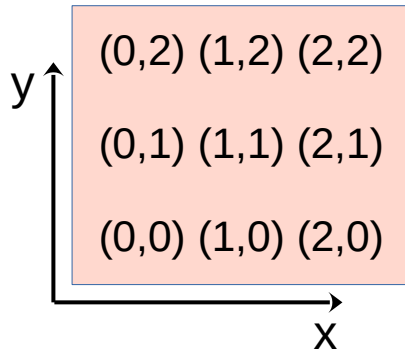
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2
0
1
2
0
1
2



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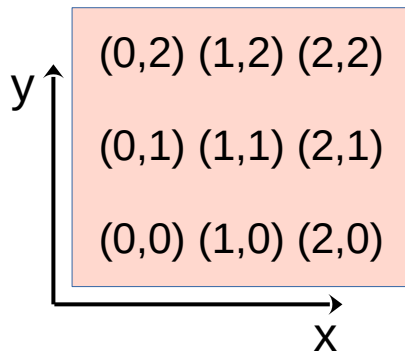
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0	0
1	0
2	0
0	1
1	1
2	1
0	2
1	2
2	2



How these python commands work? **meshgrid** and **flatten**

meshgrid: convert vectors x, y into coordinate matrices

flatten: collapses the array into one dimension

```
x = np.array([0,1,2])  
y = np.array([0,1,2])  
X, Y = np.meshgrid(x,y)
```

vectors [0, 1, 2]

Matrix:

	0	0	0
0	1	1	1
1	2	2	2

Matrix:

0	1	2
0	1	2
0	1	2

The matrix elements form the pairs $p_n = (x_n, y_n)$

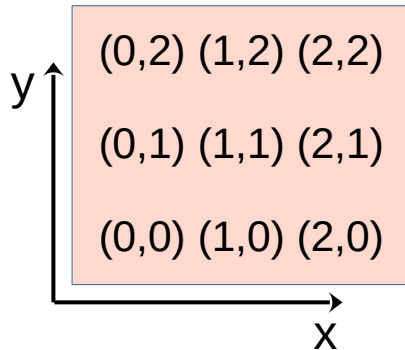
To linearize into vectors, use `.flatten()`

```
X = X.flatten()  
Y = Y.flatten()
```

0	0
1	0
2	0
0	1
1	1
2	1
0	2
1	2
2	2

$p_n = (x_n, y_n)$

(0, 0)
(1, 0)
(2, 0)
(0, 1)
(1, 1)
(2, 1)
(0, 2)
(1, 2)
(2, 2)



How these python commands work? **meshgrid** and **flatten**

meshgrid: convert vectors x, y into coordinate matrices

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y = np.array([0,1,2])  
X, Y = np.meshgrid(x,y)
```

vectors [0, 1, 2]

Matrix:

	0	1	2
0	0	1	2
1	0	1	2
2	0	1	2

Matrix:

0	0	0
1	1	1
2	2	2

↑ ↑ ↑

The matrix elements form the pairs $p_n = (x_n, y_n)$

To linearize into vectors, use `.flatten()`

```
X = X.flatten()  
Y = Y.flatten()
```

Now check this:

```
V = 0.5*(X**2 + Y**2)
```

→ is this useful? ;-)

$p_n = (x_n, y_n)$

(0, 0)
(1, 0)
(2, 0)
(0, 1)
(1, 1)
(2, 1)
(0, 2)
(1, 2)
(2, 2)



How these python commands work? **meshgrid** and **flatten**

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```

vectors [0, 1, 2]

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	0	1	2
0	0	1	2
1	0	1	2
2	0	1	2

Matrix:

0	0	0
1	1	1
2	2	2

$x \rightarrow$ $y \rightarrow$

The matrix elements form the pairs $p_n = (x_n, y_n)$

To linearize into vectors, use `.flatten()`

```
X = X.flatten()  
Y = Y.flatten()
```

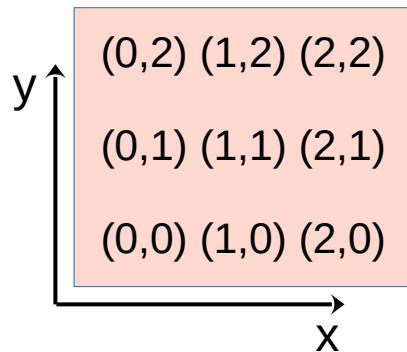
0	0
1	0
2	0
0	1
1	1
2	1
0	2
1	2
2	2

Now check this:

```
V = 0.5*(X**2 + Y**2)
```

→ is this useful? ;-)

You can also
use `kron`!
see next slides



How these python commands work? **kron**

kron: is the **Kronecker product**, or the direct product $\rightarrow \otimes$

$$p_n = (x_n, y_n)$$

(0, 0)

(1, 0)

(2, 0)

(0, 1)

(1, 1)

(2, 1)

(0, 2)

(1, 2)

(2, 2)



How these python commands work? **kron**

kron: is the **Kronecker product**, or the direct product $\rightarrow \otimes$

If A and B are matrices:

$$p_n = (x_n, y_n)$$

(0, 0)
(1, 0)
(2, 0)
(0, 1)
(1, 1)
(2, 1)
(0, 2)
(1, 2)
(2, 2)



How these python commands work? **kron**

kron: is the **Kronecker product**, or the direct product $\rightarrow \otimes$

If A and B are matrices:

$$A = \begin{pmatrix} a_{0,0} & a_{0,1} \\ a_{1,0} & a_{1,1} \end{pmatrix} \rightarrow A \otimes B = \begin{pmatrix} a_{0,0}B & a_{0,1}B \\ a_{1,0}B & a_{1,1}B \end{pmatrix}$$

$p_n = (x_n, y_n)$

(0, 0)
(1, 0)
(2, 0)
(0, 1)
(1, 1)
(2, 1)
(0, 2)
(1, 2)
(2, 2)



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Consider now the 1D matrix representations for these operators

$p_n = (x_n, y_n)$

(0, 0)
(1, 0)
(2, 0)
(0, 1)
(1, 1)
(2, 1)
(0, 2)
(1, 2)
(2, 2)



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Consider now the 1D matrix representations for these operators
(see previous classes)

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(0, 0)
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Consider now the 1D matrix representations for these operators
(see previous classes)

$$D_x = \partial_x^2$$

$$I_x = \text{identity } N_x \times N_x$$

$$p_n = (x_n, y_n)$$

(0, 0)
(1, 0)
(2, 0)
(0, 1)
(1, 1)
(2, 1)
(0, 2)
(1, 2)
(2, 2)



How these python commands work? **kron**

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Consider now the 1D matrix representations for these operators
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$$D_x = \partial_x^2$$

$$I_x = \text{identity } N_x \times N_x$$

$$D_y = \partial_y^2$$

$$I_y = \text{identity } N_y \times N_y$$

$p_n = (x_n, y_n)$

(0, 0)
(1, 0)
(2, 0)
(0, 1)
(1, 1)
(2, 1)
(0, 2)
(1, 2)
(2, 2)



How these python commands work? **kron**

kron: is the **Kronecker product**, or the direct product $\rightarrow \otimes$

If A and B are matrices:

$$A = \begin{pmatrix} a_{0,0} & a_{0,1} \\ a_{1,0} & a_{1,1} \end{pmatrix} \rightarrow A \otimes B = \begin{pmatrix} a_{0,0}B & a_{0,1}B \\ a_{1,0}B & a_{1,1}B \end{pmatrix}$$

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$$I_x = \text{identity } N_x \times N_x$$

$$D_y = \partial_y^2$$

$$I_y = \text{identity } N_y \times N_y$$

The 2D version becomes:

$$p_n = (x_n, y_n)$$

$$(0, 0)$$

$$(1, 0)$$

$$(2, 0)$$

$$(0, 1)$$

$$(1, 1)$$

$$(2, 1)$$

$$(0, 2)$$

$$(1, 2)$$

$$(2, 2)$$



How these python commands work? **kron**

kron: is the **Kronecker product**, or the direct product $\rightarrow \otimes$

If A and B are matrices:

$$A = \begin{pmatrix} a_{0,0} & a_{0,1} \\ a_{1,0} & a_{1,1} \end{pmatrix} \rightarrow A \otimes B = \begin{pmatrix} a_{0,0}B & a_{0,1}B \\ a_{1,0}B & a_{1,1}B \end{pmatrix}$$

Consider now the 1D matrix representations for these operators
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$$D_x = \partial_x^2$$

$$I_x = \text{identity } N_x \times N_x$$

$$D_y = \partial_y^2$$

$$I_y = \text{identity } N_y \times N_y$$

The 2D version becomes:

```
Dx2D = np.kron(np.eye(Ny), Dx)  
Dy2D = np.kron(Dy, np.eye(Nx))
```

$p_n = (x_n, y_n)$

(0, 0)
(1, 0)
(2, 0)
(0, 1)
(1, 1)
(2, 1)
(0, 2)
(1, 2)
(2, 2)



How these python commands work? **kron**

kron: is the **Kronecker product**, or the direct product $\rightarrow \otimes$

If A and B are matrices:

$$A = \begin{pmatrix} a_{0,0} & a_{0,1} \\ a_{1,0} & a_{1,1} \end{pmatrix} \rightarrow A \otimes B = \begin{pmatrix} a_{0,0}B & a_{0,1}B \\ a_{1,0}B & a_{1,1}B \end{pmatrix}$$

Consider now the 1D matrix representations for these operators
(see previous classes)

$$D_x = \partial_x^2$$

$$I_x = \text{identity } N_x \times N_x$$

$$D_y = \partial_y^2$$

$$I_y = \text{identity } N_y \times N_y$$

The 2D version becomes:

```
Dx2D = np.kron(np.eye(Ny), Dx)
Dy2D = np.kron(Dy, np.eye(Nx))
```

So we get:

```
H = -0.5*(Dx2D + Dy2D) + V
```

$p_n = (x_n, y_n)$

(0, 0)
(1, 0)
(2, 0)
(0, 1)
(1, 1)
(2, 1)
(0, 2)
(1, 2)
(2, 2)



Using kron instead of meshgrid and flatten

Check that these codes return the same vectors:



Using meshgrid/flatten

```
x = np.array([0,1,2])
y = np.array([3,4,5,6])
X, Y = np.meshgrid(x, y)
X = X.flatten()
Y = Y.flatten()
print('X=', X)
print('Y=', Y)
```

```
X= [0 1 2 0 1 2 0 1 2 0 1 2]
Y= [3 3 3 4 4 4 5 5 5 6 6 6]
```

Using kron instead of meshgrid and flatten

Check that these codes return the same vectors:



Using meshgrid/flatten

```
x = np.array([0,1,2])
y = np.array([3,4,5,6])
X, Y = np.meshgrid(x, y)
X = X.flatten()
Y = Y.flatten()
print('X=', X)
print('Y=', Y)
```

```
X= [0 1 2 0 1 2 0 1 2 0 1 2]
Y= [3 3 3 4 4 4 5 5 5 6 6 6]
```

Using kron

```
x = np.array([0,1,2])
y = np.array([3,4,5,6])
X = np.kron(np.ones_like(y), x)
Y = np.kron(y, np.ones_like(x))
print('X=', X)
print('Y=', Y)
```

```
X= [0 1 2 0 1 2 0 1 2 0 1 2]
Y= [3 3 3 4 4 4 5 5 5 6 6 6]
```

Results for 2D atoms or molecules

Atomic potential = inverse Gaussian

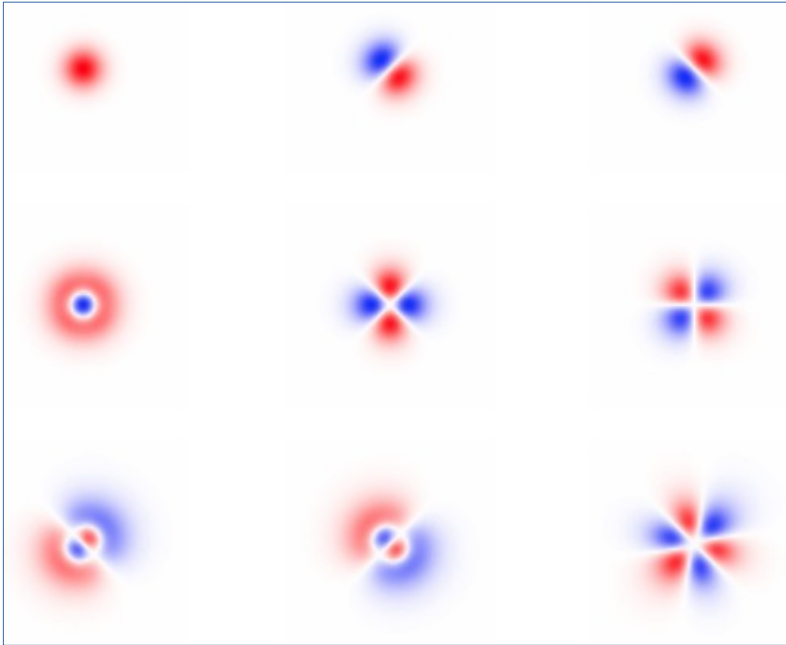


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Results for 2D atoms or molecules

Atomic potential = inverse Gaussian

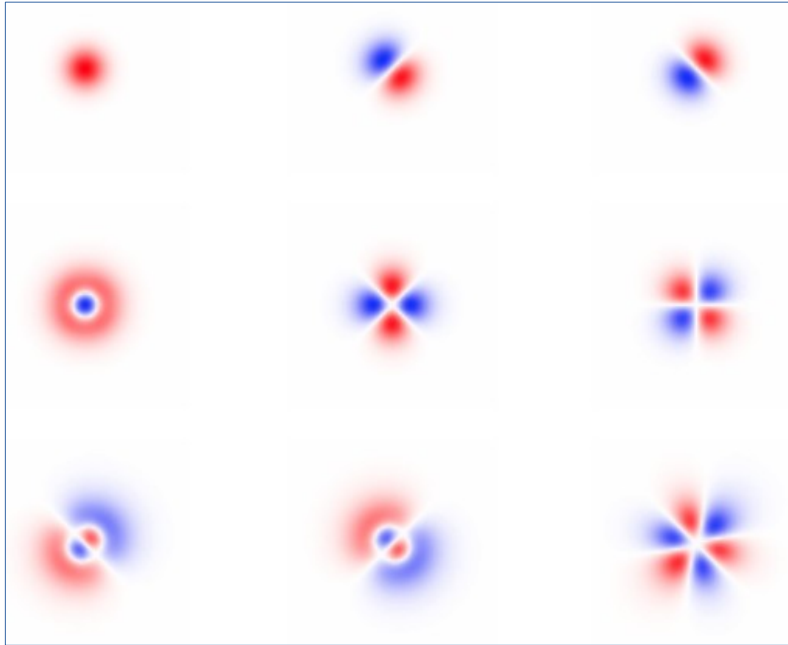
1 atom



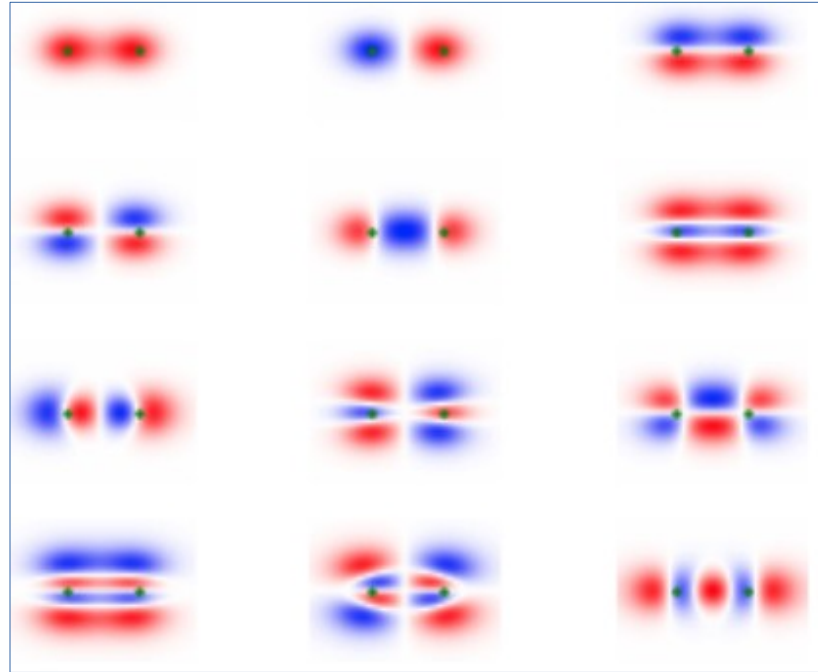
Results for 2D atoms or molecules

Atomic potential = inverse Gaussian

1 atom

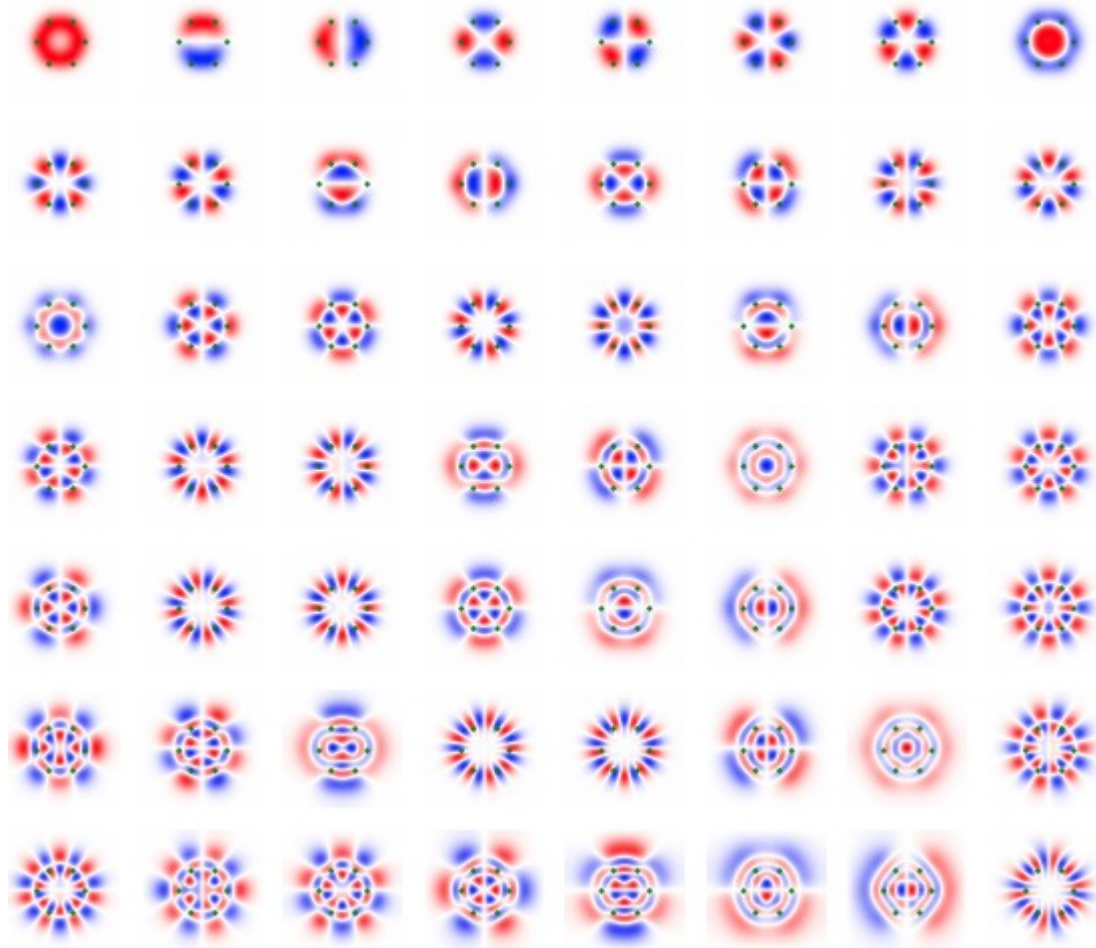


2 atoms



Results for 2D atoms or molecules

6 atoms
hexagon



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How to improve? For large matrices: use scipy sparse matrices



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How to improve? For large matrices: use scipy sparse matrices



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Sparse matrices (scipy.sparse)

SciPy 2-D sparse matrix package for numeric data.



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Sparse matrices (scipy.sparse)

SciPy 2-D sparse matrix package for numeric data.

Functions

Building sparse matrices:

`eye(m[, n, k, dtype, format])`

`identity(n[, dtype, format])`

`kron(A, B[, format])`

`kronsum(A, B[, format])`

`diags(diagonals[, offsets, shape, format, dtype])`

`spdiags(data, diags, m, n[, format])`

`block_diag(mats[, format, dtype])`



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Sparse linear algebra (scipy.sparse.linalg)



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[SciPy.org](#)[Docs](#)[SciPy v1.6.2 Reference Guide](#)

Sparse linear algebra (scipy.sparse.linalg)

Matrix factorizations

Eigenvalue problems:

`eigs(A[, k, M, sigma, which, v0, ncv, ...])`

`eigsh(A[, k, M, sigma, which, v0, ncv, ...])`



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