



## Relatório atividade 3

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# Objetivos

- Buscar uma matriz esparsa simétrica positiva definida
- Aplicar o método lanczos pelo ARPACK para e calcular autopares da matriz
- Comparar tempo de processamento para diferentes estruturas de matrizes esparsas

# Ambiente de execução

- Super computador SDumont
- Executado na fila gdl (IA)
- CPU: 2x Intel(R) Xeon(R) Gold 6148 2.4GHz
- 384 Gbs de memória ram

# Buscar matriz no ssgetpy

```
consulta = ssgetpy.search(rowbounds=(500_000,700_000),
                          colbounds=(500_000,700_000),
                          nzbounds = (1_000_000,10_000_000),
                          limit = 10)
```

consulta

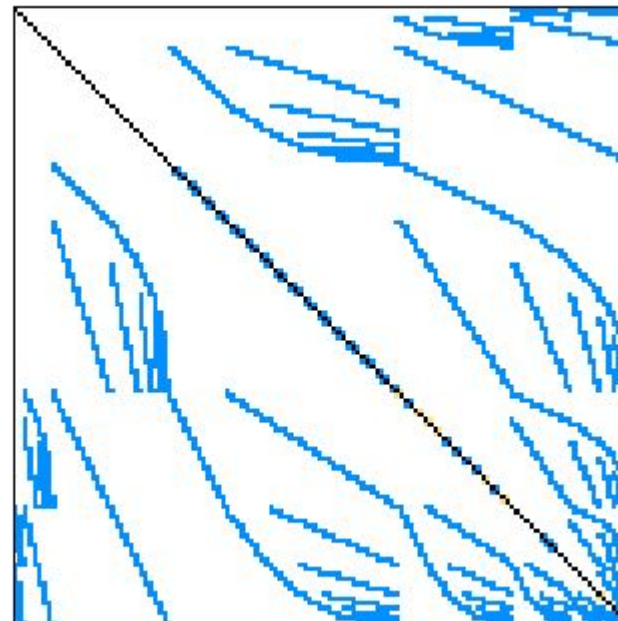
Id	Group	Name	Rows	Cols	NNZ	DType	2D/3D Discretization?	SPD?	Pattern Symmetry	Numerical Symmetry	Kind	Spy Pl
285	<a href="#">ATandT</a>	<a href="#">pre2</a>	659033	659033	5834044	real	No	No	0.33	0.065	frequency-domain circuit simulation problem	
980	<a href="#">Kamvar</a> <a href="#">Stanford Berkeley</a>		683446	683446	7583376	binary	No	No	0.25	0.25	directed graph	
1383	<a href="#">Andrianov</a>	<a href="#">lp1</a>	534388	534388	1643420	binary	No	No	1.0	1.0	optimization problem	
1396	<a href="#">Rajat</a>	<a href="#">rajat29</a>	643994	643994	3760246	real	No	No	0.69	0.69	circuit simulation problem	

```
consulta.download(destpath = f'{os.getcwd()}\\matrix',extract=True)[8]
```

# Matriz escolhida

Id: 1853   Grupo: Wissgott   Nome: parabolic\_fem

- Matriz com 525.825 linhas e colunas
- Tipo Real, com 3.674.625 números diferentes de 0
- Origem : Problema de dinâmica de fluidos computacional



# Leitura da matriz

- Formato mtx

## scipy.io.mmread

`scipy.io.mmread(source)`

[\[source\]](#)

Reads the contents of a Matrix Market file-like 'source' into a matrix.

**Parameters:** **source** : *str or file-like*

Matrix Market filename (extensions .mtx, .mtz.gz) or open file-like object.

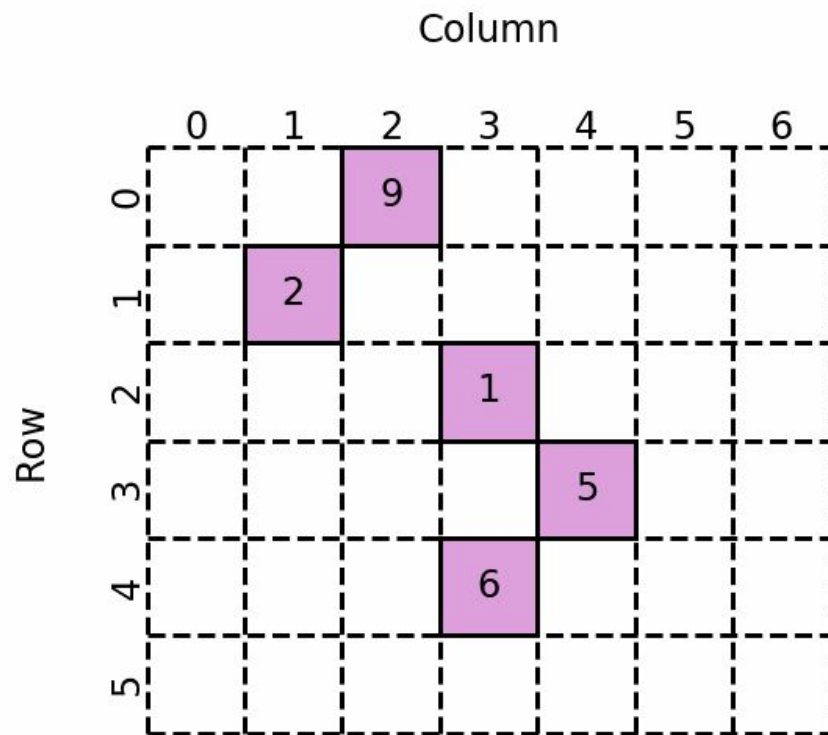
**Returns:** **a** : *ndarray or coo\_matrix*

Dense or sparse matrix depending on the matrix format in the Matrix Market file.

```
A = mmread('matrix\\parabolic_fem\\parabolic_fem.mtx')
```

```
A
```

```
<525825x525825 sparse matrix of type '<class 'numpy.float64'>'  
  with 3674625 stored elements in COOrdinate format>
```



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Fonte: <https://develloppaper.com/instructions-for-using-python-scipy-sparse-matrix/>

# COO

Row

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

Column

1	4	2	3	3
---	---	---	---	---

Data

2	5	9	1	6
---	---	---	---	---

## scipy.sparse.csr\_matrix

```
class scipy.sparse.csr_matrix(arg1, shape=None, dtype=None, copy=False)
```

Compressed Sparse Row matrix

This can be instantiated in several ways:

**csr\_matrix(S)**

with another sparse matrix S (equivalent to S.tocsr())

```
A2 = csr_matrix(A)
A2
```

```
<525825x525825 sparse matrix of type '<class 'numpy.float64'>'
  with 3674625 stored elements in Compressed Sparse Row format>
```

## scipy.sparse.csc\_matrix

```
class scipy.sparse.csc_matrix(arg1, shape=None, dtype=None, copy=False)
```

Compressed Sparse Column matrix

This can be instantiated in several ways:

**csc\_matrix(S)**

with another sparse matrix S (equivalent to S.tocsc())

```
A3 = csc_matrix(A)
A3
```

```
<525825x525825 sparse matrix of type '<class 'numpy.float64'>'
  with 3674625 stored elements in Compressed Sparse Column format>
```



# Autopares

## scipy.sparse.linalg.eigsh

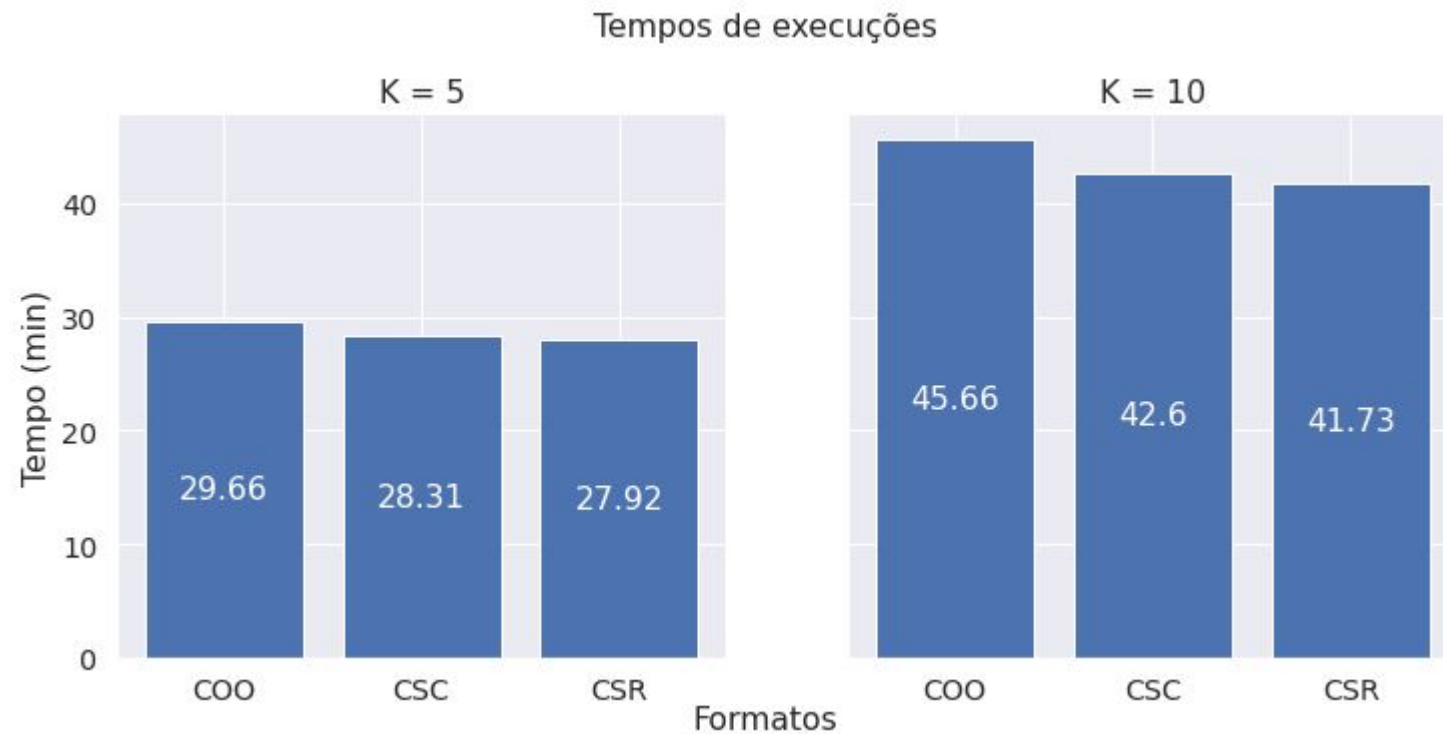
```
scipy.sparse.linalg.eigsh(A, k=6, M=None, sigma=None, which='LM', v0=None, ncv=None,  
maxiter=None, tol=0, return_eigenvectors=True, Minv=None, OPinv=None, mode='normal')
```

Find k eigenvalues and eigenvectors of the real symmetric square matrix or complex Hermitian matrix A. [\[source\]](#)

```
for n in [5,10]:  
    print(n)  
    np.random.seed(18052022)  
    v0 = np.random.rand(min(A.shape))  
    start = time.time()  
    eigenvalues, eigenvectors = eigsh(A, k=n, v0=v0)  
    end = time.time()  
    print(end-start)  
    np.save("/scratch/rtm-ug/lyncoln.oliveira/lanczos/eigenvalues_COO_"+ str(n) + ".np", eigenvalues)  
    np.save("/scratch/rtm-ug/lyncoln.oliveira/lanczos/eigenvectors_COO_"+ str(n) + ".np", eigenvectors)  
    np.save("/scratch/rtm-ug/lyncoln.oliveira/lanczos/time_COO_"+ str(n) + ".np", end-start)
```

# Resultados

- $K = 5$  : [0.79998248, 0.7999853 , 0.79998905, 0.79999375, 0.79999657]
- $K = 10$  : [0.79996651, 0.7999712 , 0.79997401, 0.79997778, 0.79998247, 0.79998248, 0.7999853 , 0.79998905, 0.79999375, 0.79999657]



# Conclusões

- Economia memória
- Obtenção dos autopares em bom tempo
- Pequena diferença entre os formatos CSR e CSC
- Mais execuções