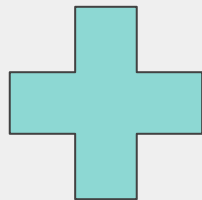


# Vettorizzazione: a hands-on approach in Python

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5
9
1
4
5
1
3



1
3
2
4
5
1
3

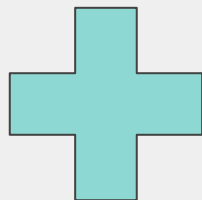




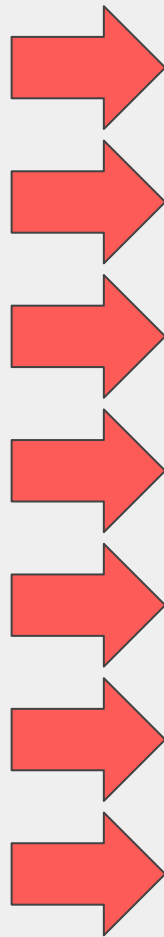
## Senza vettorizzazione

```
1. v1 = [5, 9, 1, 4, 5, 1, 3]
2. v2 = [1, 3, 2, 4, 5, 1, 3]
3.
4. vout = []
5. for i in range(len(v1)):
6.     vout.append(v1[i] + v2[i])
```

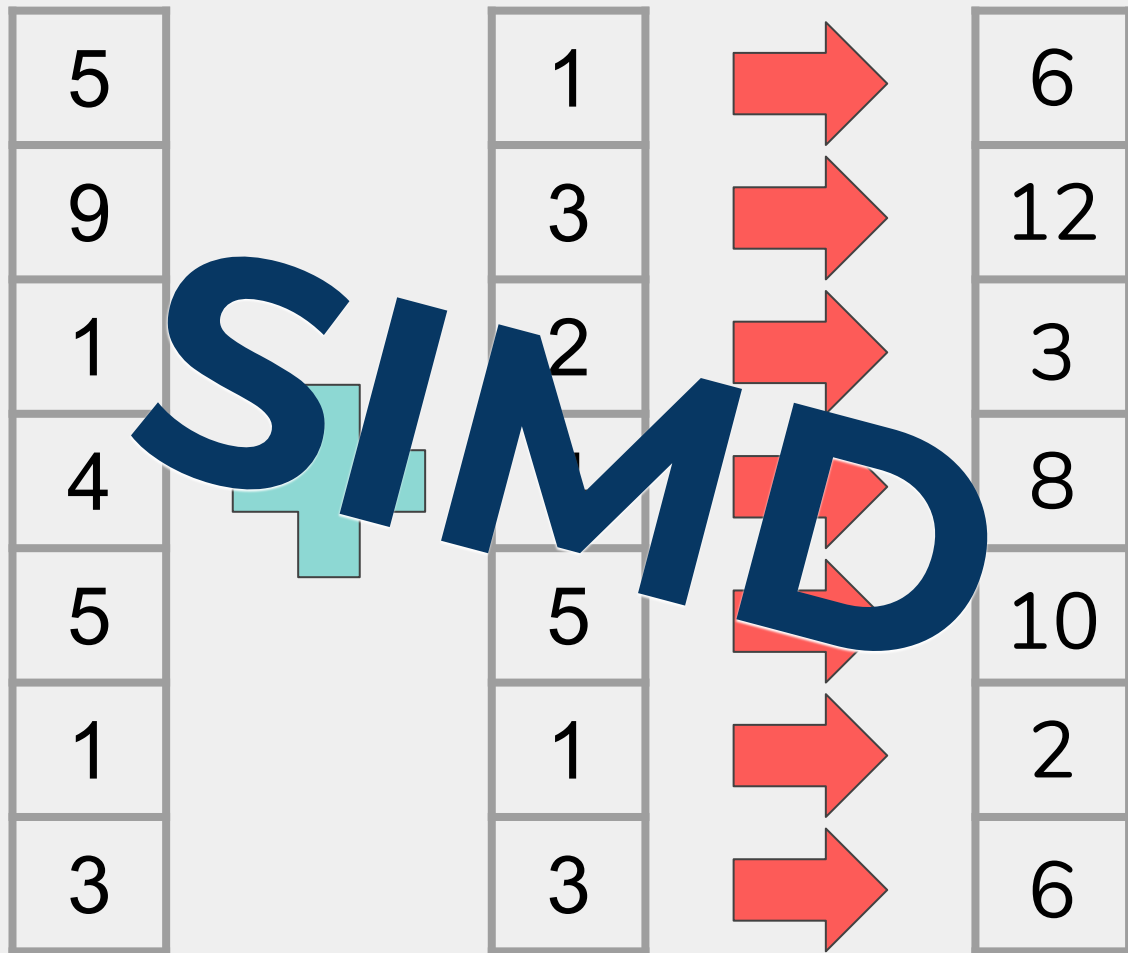
5
9
1
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6
12
3
8
10
2
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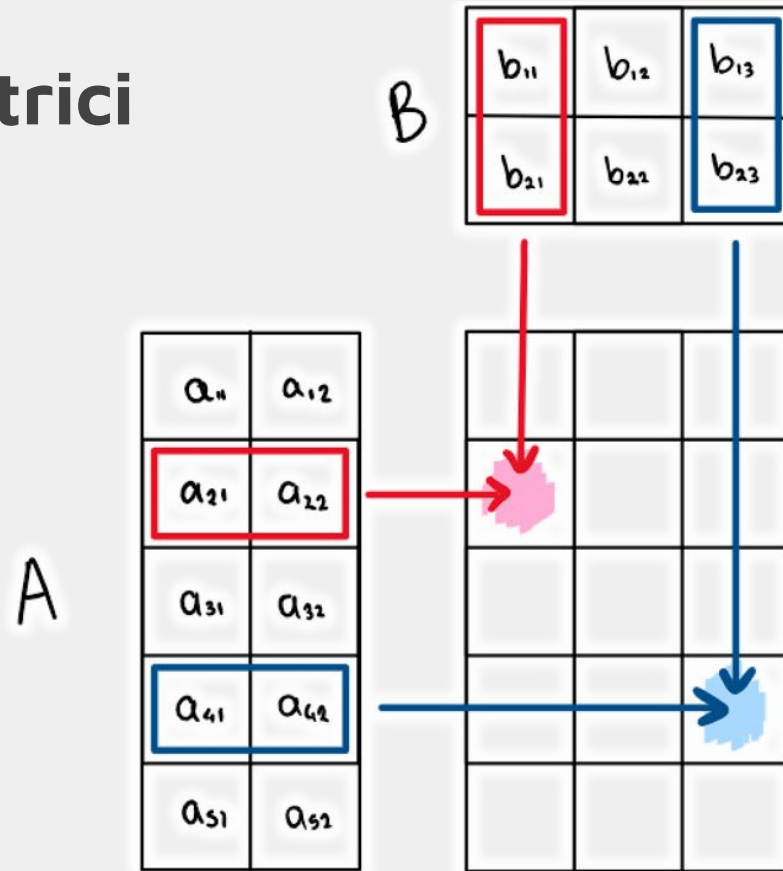


# Benchmark

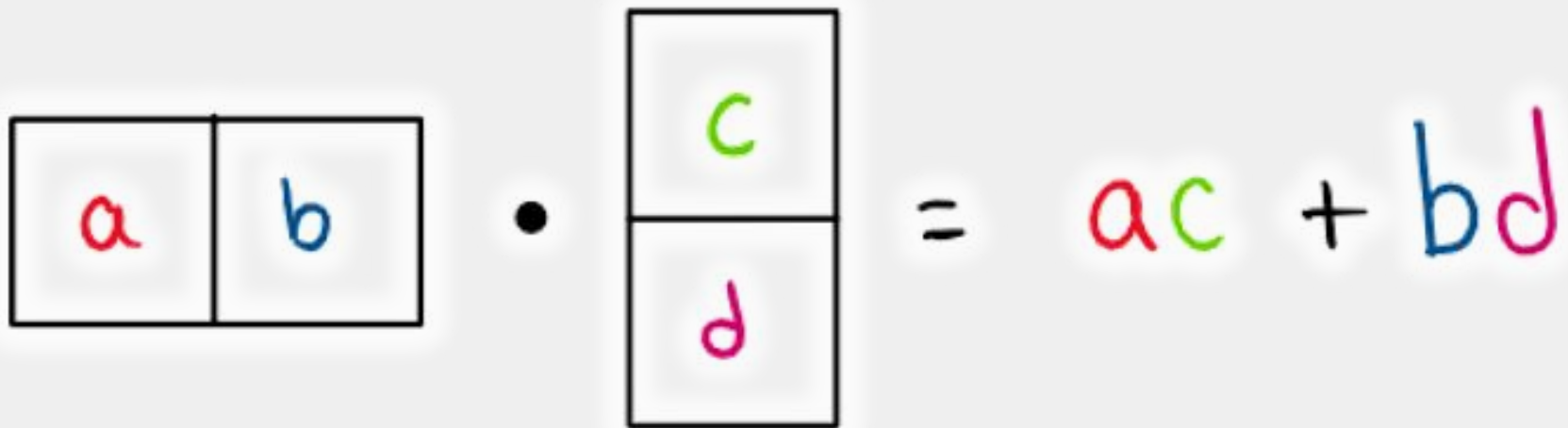


# Prodotto tra matrici

$C = A B$



## Prodotto scalare


$$\begin{bmatrix} a & b \end{bmatrix} \cdot \begin{bmatrix} c \\ d \end{bmatrix} = ac + bd$$





## Senza vettorizzazione

```
1.  def my_matmul(A, B):
2.      assert A.shape[1] == B.shape[0]
3.      C = np.zeros((A.shape[0], B.shape[1]), dtype=np.int64)
4.
5.      for i in range(C.shape[0]):
6.          for j in range(C.shape[1]):
7.              col = B[:, j] # Colonna j-esima
8.              row = A[i, :] # Riga i-esima
9.              inner_out = 0
10.             for inner in range(len(col)):
11.                 inner_out = inner_out + (col[inner] * row[inner])
12.             C[i, j] = inner_out
13.     return C
```



## Senza vettorizzazione (Cython)

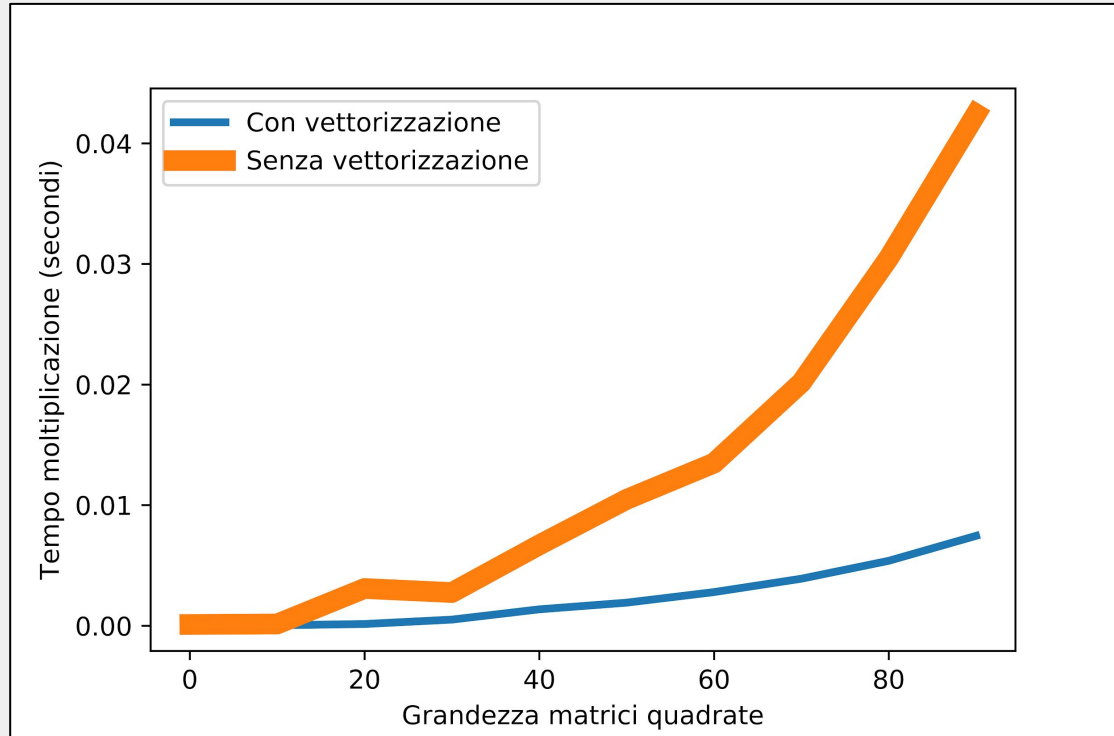
```
1.  def my_matmul_opt (long[:, ::1] A, long[:, ::1, :] B):
2.      assert A.shape[1] == B.shape[0]
3.      cdef long[:, :] C = np.zeros((A.shape[0], B.shape[1]),
dtype=np.int64)
4.
5.      cdef long inner_out = 0L
6.      for i in range(C.shape[0]):
7.          for j in range(C.shape[1]):
8.              inner_out = 0L
9.              for inner in range(A.shape[1]):
10.                  inner_out = inner_out + (B[inner, j] * A[i, inner])
11.              C[i, j] = inner_out
12.      return np.asarray(C)
```



## Con vettorizzazione

```
C = np.matmul(A, B)
```

# Confronto performance



**Quali sono le differenze  
tra vettorizzazione e  
programmazione  
multithreading?**





# **Intel Advanced Vector Extensions (AVX)**

Insieme di istruzioni per effettuare  
operazioni SIMD su processori Intel/AMD

add / sub

max / min

average

div / mul

and / or / xor / ...

dot product

bit shift

compare

... e molto altro



# Intel Advanced Vector Extensions (AVX)

- Registri fino a 512 bit (AVX-512)
- Operazioni non distruttive:  $A = B + C$
- FMA:  $a = a + (b * c)$
- CUID per il supporto HW e dell'OS
- Intel rilascia un emulatore per il testing





## Intel Xenon Phi VPU

Solitamente:

- latenza: 4 cicli
- throughput: 1 ciclo

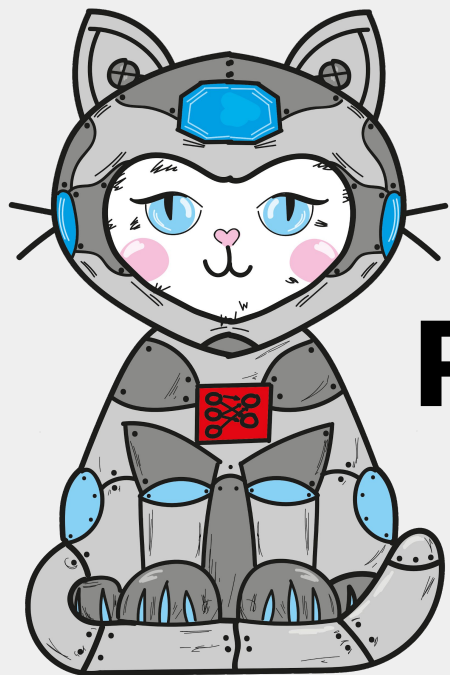


## What's next?

- Anaconda:  
<https://www.anaconda.com/distribution/>
- NumPy - Quickstart tutorial:  
<https://numpy.org/devdocs/user/quickstart.html>

# Domande





# Rage Against the Data

<https://tiny.cc/RageAgainstTheData>