

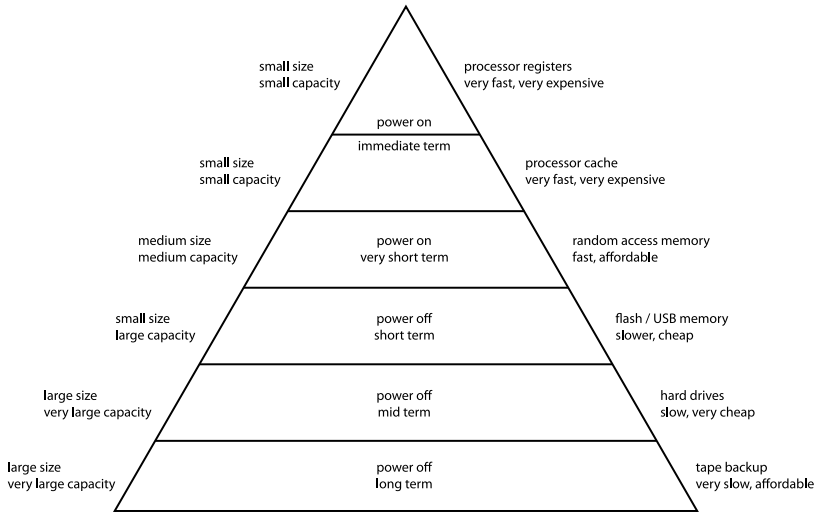
Programming (Econometrics)

Lecture 3: Memory organization

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Computer Memory Hierarchy



- Local variables such as loop counters can possibly be stored in registers
- All larger data structures have to be allocated to the main memory
- The random access memory is linear and addressed using integers pointing out the location (e.g. 0x400345CF)
- 32 bit addressing = max 4Gb of memory

- Matrices are included in Matlab as a built-in data type

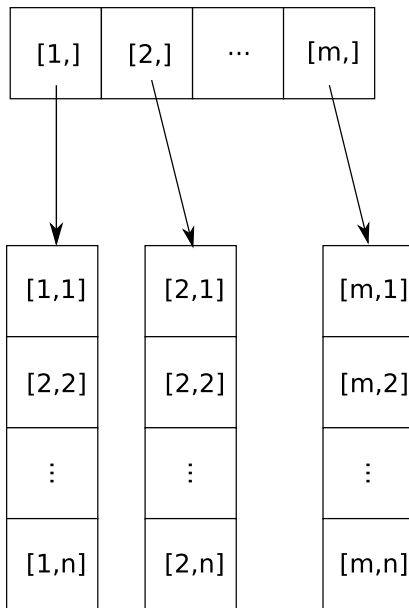
```
a = [3 , 4];
```

```
b = '1';
```

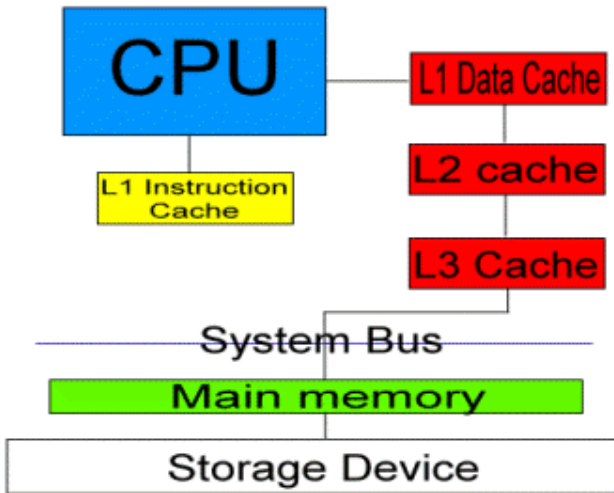
```
c = a*b; % what 's c now?
```

- How to represent $m \times n$ matrices?

Matrix representations: naive



CPU caches



Matrix representations: efficient

- Memory is linear, so store the element $[a, b]$ in index $[(a - 1) * n + b]$

1	2	...	n	(n+1)	(n+2)	...	(m*n)
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- Row-major representation; in column-major one $[a, b]$ is in $[(b - 1) * m + a]$
- In most programming languages the array indices start from 0 and the formulas are simpler

Row- and column-major representations: an example

$$\begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix}$$

- As row-major: [1 2 3 4 5 6]
- As column-major: [1 4 2 5 3 6]

Special matrices: sparse

- If the matrix is *sparse*, i.e. it contains only a few elements, it is more efficient to store only the non-zero elements
- E.g.

$$\begin{bmatrix} 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 2 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 \end{bmatrix}$$

- Can be represented with $([3, 4, 2], [5, 1, 1])$

Special matrices: diagonal and identity

$$\begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 3 & 0 & 0 & 0 \\ 0 & 0 & 2 & 0 & 0 \\ 0 & 0 & 0 & 7 & 0 \\ 0 & 0 & 0 & 0 & 4 \end{bmatrix}$$

- Can be represented with $[1, 3, 2, 7, 4]$

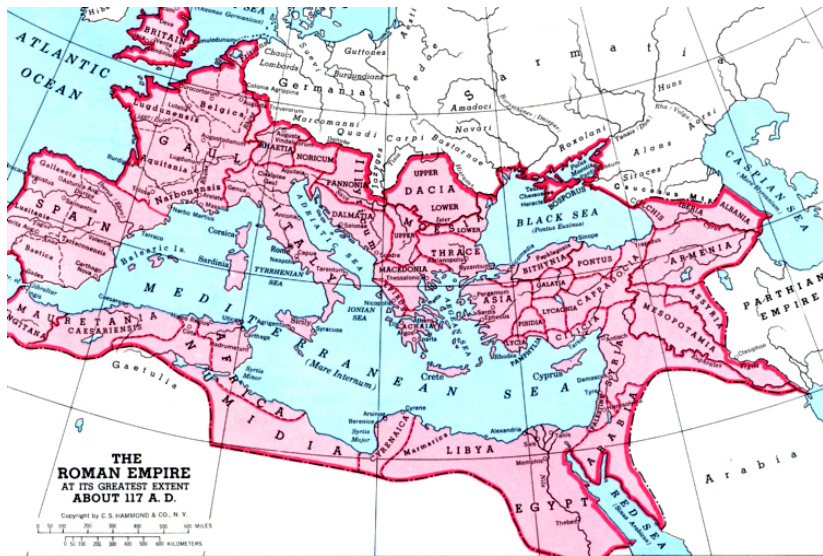
$$\begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix}$$

- $= I_5$ and can be represented with a single integer 5

Matrix multiplication: naive

```
function C = multiply(A, B)
    C = zeros(rows(A), columns(B));
    for (i=1:rows(A))
        for (j=1:columns(B))
            s = 0;
            for (k=1:columns(A))
                s = s + A(i, k) * B(k, j);
            end
            C(i, j) = s;
        end
    end
end
```

- Complexity?



Matrix multiplication: divide-and-conquer

- Assume that we are multiplying $n \times n$ matrices, where n is a power of 2
- Express $C = AB$ as

$$\begin{bmatrix} C_{1,1} & C_{1,2} \\ C_{2,1} & C_{2,2} \end{bmatrix} = \begin{bmatrix} A_{1,1} & A_{1,2} \\ A_{2,1} & A_{2,2} \end{bmatrix} \begin{bmatrix} B_{1,1} & B_{1,2} \\ B_{2,1} & B_{2,2} \end{bmatrix}$$

that comes down to computing

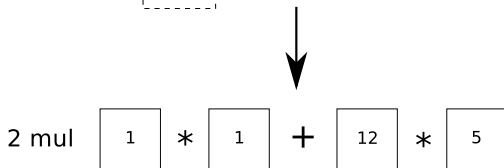
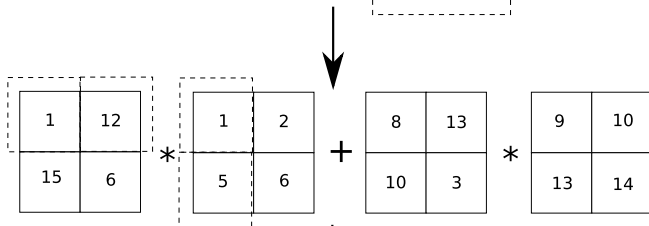
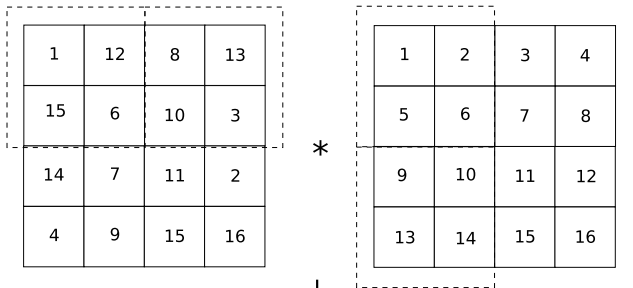
$$C_{1,1} = A_{1,1}B_{1,1} + A_{1,2}B_{2,1}$$

$$C_{1,2} = A_{1,1}B_{1,2} + A_{1,2}B_{2,2}$$

$$C_{2,1} = A_{2,1}B_{1,1} + A_{2,2}B_{2,1}$$

$$C_{2,2} = A_{2,1}B_{1,2} + A_{2,2}B_{2,2}$$

- Proceed recursively until you multiply matrices of max size 1×1



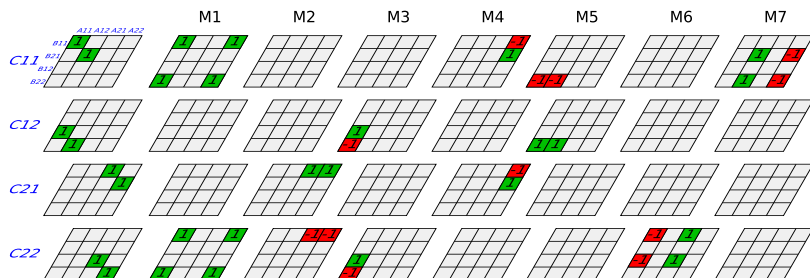
Complexity of divide-and-conquer multiplication

$$\begin{aligned}T(n) &= 8T(n/2) + n^2 \\&= n^2 + 8((n/2)^2 + 8T(n/4)) \\&= n^2 + 8((n/2)^2 + 8((n/4)^2 + 8T(n/16))) \\&= n^2 + 2n^2 + 4n^2 + 8T(n/16))\end{aligned}$$

i^{th} term in the series is $2^{i-1}n^2$

$$\begin{aligned}T(n) &= n^2 + 2n^2 + 4n^2 + \dots + 2^{\log_2 n} O(1) \\&= n^2 \sum_{i=0}^{\log_2 n} 2^i + O(n^{\log_2 2}) \\&= n^2 \frac{2^{\log_2(n+1)} - 1}{2 - 1} + O(n) \\&\leq n^2 O(2^{\log_2 n}) + O(n) = n^2 O(n) + O(n) \\&= O(n^3)\end{aligned}$$

Strassen's idea



Now we only need to do 7 multiplications, so the complexity becomes

$$\begin{aligned}
 T(n) &= 7T(n/2) + O(n^2) \\
 &= O(n^{\log_2 7}) \approx O(n^{2.81})
 \end{aligned}$$

- Matrices and arrays are static data structures in the sense that although accessing an arbitrary element is efficient, adding an element is not
- Example: add an element into an array

Complexity of operations with matrices and arrays

For n elements

- Add element: $O(n)$
- Random access: $O(1)$
- Delete element: $O(n)$

For $n \times n$ matrices:

- Multiplication: $O(n^3)$ (?)
- Inversion: as multiplication
- Determinant: $O(n^3)$ with LU decomposition