## Lecture 15: Complexity

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- Complexity of factorization algorithms
  - Complexity of QR

2 LU factorization

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### LU-factorization

• LU-factorization applies to  $A m \times m$ :

$$A = LU$$
.

Here, L is lower triangular, and U is upper triangular.

- LU-factorization is not used to decompose rectangular matrices, or to solve least squares.
- It is used in order to solve square, invertible systems.

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### LU-factorization

Q: QR- or LU-? Which method to prefer?

#### We will compare the two methods based on:

- Complexity
- Stability: To recall, whenever A = QR,

$$cond(A) = cond(Q)cond(R),$$

which is best possible. However, for LU, it is possible to have, if A=LU,

$$cond(A) << cond(L)cond(U)$$
.

 So, unless LU has better complexity, there is no motivation to use, not even to learn it!!

### LU-factorization

Q: QR- or LU-? Which method to prefer? In order to compare the complexity of QR and LU we need

- To learn LU-factorization...
- To understand the complexity of LU.
- To understand the complexity of QR.

#### A square, A = QR. Output of QR is

- The matrix R.
- The vectors  $w_1, \ldots, w_{m-1}$  that make the matrices  $H_1, \ldots, H_{m-1}$ .
- We do not compute the Householder matrices, let alone *Q*.

#### How to solve then the linear system Ax = b?

We need to solve

$$Rx = Q'b$$
,

so we need to know Q'b without computing Q.

•  $Q = H_1 \dots H_{m-1}$ , hence

$$Q'b = H_{m-1}(....H_2(H_1b))....).$$

• So, we just need to know how to compute Hv, when H = I - 2ww'.

#### The key computation:

- Input: Vectors  $v, w \in \mathbb{R}^m$ , such  $w(i) = 0, 1 \le i \le k$ .
- Output: The vector (I 2ww')v.

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$$(I - 2ww')v = v - 2w(w'v) = v - 2(v, w)w.$$

#### Counting multiplication:

- Computing (v, w): m k.
- Computing 2(v, w): m k + 1.
- Computing 2(v, w)w:  $2(m k) + 1 \approx 2(m k)$ .

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- Computing (v, w): m k.
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- Computing 2(v, w)w:  $2(m k) + 1 \approx 2(m k)$ .

In Step I,  $w_1$  has not zeros (k = 0).

#### We need to

- Compute  $w_1$  This will require O(m) operations.
- Compute  $(I 2w_1w_1')A_0$ : Need to update m 1 columns in  $A_0$ .
- Investment: 2(m-0)(m-1) = 2m(m-1)

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Counting multiplication:

- Computing (v, w): m k.
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In Step II,  $w_2$  has one zero (k = 1). We need to

- Compute  $w_2$  This will require O(m) operations.
- Compute  $(I 2w_2w_2')A_1$ : Need to update m 2 columns in  $A_1$ .
- Investment: 2(m-1)(m-2)

So, total investment is

$$2(m(m-1)+(m-1)(m-2)+\ldots+2\times 1)=\frac{2(m-1)m(m+1)}{3}.$$

# The complexity of QR-factorization

#### F

or a matrix of order m, efficient QR-factorization consumes  $2m^3/3$  multiplications (+ a few lower order terms).

We will see later that LU-factorization consumes  $m^3/3$  operations.

- Complexity of factorization algorithms
  - Complexity of QR

2 LU factorization

#### Goal:

Factor  $A m \times m$  into A = LU, where

- L is lower triangular
- U is upper triangular

#### Process:

Create a sequence of matrices

$$A = A_0 \mapsto A_1 \mapsto \dots A_{m-1} = U.$$

Here:

$$A_j = L_j A_{j-1},$$

with  $L_i$  lower triangular. Then

$$U=L_{m-1}L_{m-2}\ldots L_1A,$$

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$$L_j = I - l_j e_j',$$

with  $l_j(i) = 0$ , for  $i \le j$ .

If 
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Proof:

$$(I - le'_i)(I + le'_i) = I - le'_i + le'_i - le'_i le'_i = I - le'_i le'_i.$$

However:

## If $B = I - le'_i$ , where l(i) = 0, $i \le j$ , then

$$B^{-1} = I + le_j'.$$

Proof:

$$(I - le'_i)(I + le'_i) = I - le'_i + le'_i - le'_i le'_i = I - le'_i le'_i.$$

However:

$$le'_{j}le'_{j} = l(e'_{j}l)e'_{j} = l(j)le'_{j} = 0.$$

# How to compute the L-matrix?

We have that

$$U = L_{m-1} \dots L_1 A$$
.

Therefore,

$$A = L_1^{-1} \dots L_{m-1}^{-1} U.$$

We have just observed that inverting  $L_j$  is trivial.

But: How to multiply the inverses?

## How to compute the L-matrix?

If 
$$B_j = I + l_j e'_j$$
, where  $l_j(i) = 0$ ,  $i \le j$ , then

$$B_1B_2...B_{m-1} = I + l_1e'_1 + l_2e'_2 + ...l_{m-1}e'_{m-1}.$$

## How to compute the L-matrix?

### If $B_j = I + l_j e'_j$ , where $l_j(i) = 0$ , $i \leq j$ , then

$$B_1B_2...B_{m-1} = I + l_1e'_1 + l_2e'_2 + ...l_{m-1}e'_{m-1}.$$

Proof: When expanding, every term other than those listed above contains a sequence

$$l_j e'_j l_k e'_k$$

where j < k. But:

$$l_j e'_j l_k e'_k = l_j (e'_j l_k) e'_k = l_k (j) l_j e'_k = 0.$$