6.: 30.00
$$\chi_1$$
 + 591400 χ_2 = 591700
E2: 5.291 χ_1 - 6.130 χ_2 = 46.78, USING
4 digit precision:
When = $\frac{5.291}{30.00}$ = 0.1764, then E2-MarE_1 = E2:
 $\int 36.00 \chi_1 + 5(1400 \chi_2 = 551700)$
 $-104300 \chi_2 = -104400$
 $\chi_1 = -10$ $\chi_2 = 1.001$ (wrong!)
Problem is the disponenticises of quantities...
A simple free iz to use total or partial scaling:
The simplest is rowscaling.¹
 $\chi_{\chi} = b$
Multiply E; by Vi , i=1,7...,m, or
 $D_1 \Lambda \chi = D_1 D_1 = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$

lef C=Dià churse ri sother C={Gij} satisfies max [c;j]~1 i=1,2...,m So $\frac{L}{\Gamma_{i}} = \max \left\{ \begin{array}{c} a_{ij} \\ \vdots \\ \end{array} \right\}$ c? $\therefore \quad C_{ij} = r_i a_{ij} \qquad i = l_j 2 - m_j m_j$ This might not be helpful (need to check the condition number of Cx=D, b (drissen) and check each row for loss precision in the subtraction stage Scaling can be dere once, before the GE, or done as GE proceeds (in both cases, you still need to worry about stability) Scaling Ax=b is usually a good idea regardless of numerical litear algebra solver strategy used , i.e. GE, BR, SVD, etc

Runk: Noteven Ax=b can be LU-solved.
ex)
$$4x_2 = 8$$

 $4x_1 + 2x_2 = 17$
 $A = \begin{bmatrix} 0 \\ 42 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \begin{bmatrix} u_{11} \\ u_{12} \end{bmatrix} \begin{bmatrix} u_{11} \\ 0 \\ u_{22} \end{bmatrix}$
 $u_{11} = 0$
 $a_{21}u_{11} = 4$ (not possible)?
However,
 $4x_1 + 2x_2 = 17$
 $4x_2 = 8$ can be LU deconposed:
 $\begin{bmatrix} 4 \\ 2 \\ 0 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \begin{bmatrix} 4 \\ 2 \\ 0 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \begin{bmatrix} 4 \\ 2 \\ 0 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \begin{bmatrix} 4 \\ 2 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \end{bmatrix} \begin{bmatrix} 2 \\ 0 \end{bmatrix} \begin{bmatrix} 4 \\ 2 \\ 0 \end{bmatrix}$
Runk: For $5x = b$, if b is Diabonally-Donument
GE can be performed w/b row exchapes.
 $X = \{a_{ij}\}$ is Diabonally-Donument if
 $|a_{ii}| > \sum_{i=1}^{m} [a_{ij}] = 0, 1, ..., m$
 $i \neq i$

Lecture 23 Cholesky Factorischen It's an LU factorization, applicable to (HPD) HERHITIAN POSITIVE DEFINITOVE MATRICES if AE CTAN IF DETROMANN HPD is SPD (Symmetric Positive Définite matrix. def: DETRMAN is symmetric if N=N (aij=aji) def: AE (Mxm is Hernitian of 15 = aju) (noter this news that the diagonal entries worst be real) IP & is Hermitian Hen $\chi^* A y = y^* A \chi \quad \forall x, y \in \mathbb{C}^m$ holds. In particular, if y=x and xEIm x* & is real If, in addition,

x*1x>0 Vx=0 A is HPD

SOME PROPERTIES OF HERMITIAN & HPD :

* He e'velues of NPD one > O, and Real (in fact if you can show that & has all positive recl evelues and is symmetric => A is NPD) * The electors of Hermitic native corresponding to distinct evelues are L. * IF A has m évolues the A=QAQ* and elvectors are I and REOL Solving Ax= b, vorz Cholesky Cholesky is a special variant of LU, i.e. you set le same decomposition. Note the LU deunposition of an arbitrary & is not unique. However iFA is symptoic the LU (Cholesky) will be migue. We'll see why in what follows: Let AE CMXM HPD

Perform let step of GE:
let
$$z = \sqrt{a_1}$$
 (F)
 $A_{j} \begin{bmatrix} a_{11} \begin{bmatrix} w^{*} \\ W \end{bmatrix} \end{bmatrix} = \begin{bmatrix} a_{2} \\ Wa \end{bmatrix} = \begin{bmatrix} a_{11} \begin{bmatrix} w^{*} & 1 \\ W \end{bmatrix} \end{bmatrix} = \begin{bmatrix} a_{11} \begin{bmatrix} w^{*} & 1 \\ W \end{bmatrix} \end{bmatrix} = \begin{bmatrix} a_{11} \begin{bmatrix} a_{11} & w^{*} \\ W \end{bmatrix} \end{bmatrix} = \begin{bmatrix} a_{11} \begin{bmatrix} a_{11} & w^{*} \\ W \end{bmatrix} \end{bmatrix} = \begin{bmatrix} a_{11} \begin{bmatrix} a_{11} & w^{*} \\ W \end{bmatrix} \end{bmatrix} = \begin{bmatrix} a_{11} \begin{bmatrix} a_{11} & w^{*} \\ W \end{bmatrix} \end{bmatrix} = \begin{bmatrix} a_{11} \begin{bmatrix} a_{11} & w^{*} \\ W \end{bmatrix} \end{bmatrix} = \begin{bmatrix} a_{11} \begin{bmatrix} a_{11} & w^{*} \\ W \end{bmatrix} \end{bmatrix} = \begin{bmatrix} a_{11} \begin{bmatrix} a_{11} & w^{*} \\ W \end{bmatrix} = \begin{bmatrix} a_{11} \begin{bmatrix} w^{*} \\ W \end{bmatrix} \end{bmatrix} = 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\end{bmatrix} = \begin{bmatrix} a_{12} \begin{bmatrix} a_{12}$

and recursively repeat process in times. In the end
the middle metrix will be Imam . This preventes
a factorization: Cholesky factorization
$$A = R^*R = R_1^*R_2^* \cdots R_m^*R_{m-1} \cdots R_2R_1$$

 $R^* \qquad R$

(j)>0 Note: We assured m (2) that an >0 and used this, or swelling silver for each decopysition stage. They most be >0, since K-WWK/a is positive definite, then we've OK. Thim: Every HPD se Conton has a unique Cholesky decomposition. Pt: see algoridhm. Conputationel Complexity: for RECMAN Storage is O(mil) and the speed of algorithmiss O(m3) flops.

Stability: Ar=6 Choleshy is backwend stable.