

V vector space $\xrightarrow{L}$ W vector space
 S basis of V T basis of W

Look for a matrix M_{TS}^L so that the mapping L is represented via matrix multiplication.

Fact: $W; T \xrightarrow{K} X; U$

$$(K \circ L)(x) = K(L(x))$$

then

$$M_{US}^{K \circ L} = M_{UT}^K \cdot M_{TS}^L$$

↑
need to match

Example: Bernstein Polynomials:

$$1 = x + (1-x) = (x + (1-x))^n = \sum_{i=0}^n \binom{n}{i} x^i (1-x)^{n-i}$$

$$\frac{n!}{i!(n-i)!}$$

$$n! = 1 \cdot 2 \cdot \dots \cdot n$$

For $n=2$:

$$1 = (x + (1-x))^2 = \underbrace{x^2}_{B_2^2} + \underbrace{2x(1-x)}_{B_1^2} + \underbrace{(1-x)^2}_{B_0^2}$$

$$S = [B_0^2, B_1^2, B_2^2]$$

$$T = [1, x, x^2]$$

$$L: \mathbb{F}[x, 2] \rightarrow \mathbb{F}[x, 2]$$

$$p \mapsto p'$$

$$M_{TT}^L = \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 2 \\ 0 & 0 & 0 \end{pmatrix}$$

$$\begin{aligned} L(1) &= 1' = 0 && \leftarrow 1^{\text{st}} \text{ column} \\ L(x) &= x' = 1 = 1 \cdot 1 + 0 \cdot x + 0 \cdot x^2 \\ L(x^2) &= (x^2)' = 2x \\ &= 0 \cdot 1 + 2 \cdot x + 0 \cdot x^2 \end{aligned}$$

Goal: Write out $M_{SS}^L = M_{ST}^{\text{Id}} M_{TT}^L M_{TS}^{\text{Id}}$

$$= \left(M_{TS}^{\text{Id}} \right)^{-1}$$

$$B_0^2 = (1-x)^2 = 1 - 2x + x^2$$

$$B_1^2 = 2x(1-x) = 2x - 2x^2$$

$$B_2^2 = x^2$$

$$M_{TS}^{\text{Id}} = \begin{pmatrix} 1 & 0 & 0 \\ -2 & 2 & 0 \\ 1 & -2 & 1 \end{pmatrix}$$

$$\left(\begin{array}{ccc|ccc} 1 & 0 & 0 & 1 & 0 & 0 \\ -2 & 2 & 0 & 0 & 1 & 0 \\ 1 & -2 & 1 & 0 & 0 & 1 \end{array} \right) \rightarrow \left(\begin{array}{ccc|ccc} 1 & 0 & 0 & 1 & 0 & 0 \\ 0 & 2 & 0 & 2 & 1 & 0 \\ 0 & -2 & 1 & -1 & 0 & 1 \end{array} \right)$$

$$\Rightarrow M_{ST}^{Id} = (M_{TS}^{Id})^{-1} = \begin{pmatrix} 1 & 0 & 0 \\ 1 & \frac{1}{2} & 0 \\ 1 & 1 & 1 \end{pmatrix}$$

$$\bullet \quad M_{TT}^L M_{TS}^{Id} = \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 2 \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ -2 & 2 & 0 \\ 1 & -2 & 1 \end{pmatrix} = \begin{pmatrix} -2 & 2 & 0 \\ 2 & -4 & 2 \\ 0 & 0 & 0 \end{pmatrix} = M_{TS}^L$$

$$\bullet \quad M_{SS}^L = M_{ST}^{Id} M_{TS}^L = \begin{pmatrix} 1 & 0 & 0 \\ 1 & \frac{1}{2} & 0 \\ 1 & 1 & 1 \end{pmatrix} \begin{pmatrix} -2 & 2 & 0 \\ 2 & -4 & 2 \\ 0 & 0 & 0 \end{pmatrix} = \begin{pmatrix} -2 & 2 & 0 \\ -1 & 0 & 1 \\ 0 & -2 & 2 \end{pmatrix}$$

Check: $(B_0^2)'$ ^{not direct} $\downarrow$ $= -2B_0^2 - 1B_1^2$

$\nwarrow$ represents $M_{SS}^L \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} = \begin{pmatrix} -2 \\ -1 \\ 0 \end{pmatrix}$ represents $-2B_0^2 - 1B_1^2 = -2(1-2x+x^2) - (2x-2x^2)$

$= -2 + 4x - 2x^2 - 2x + 2x^2$

$= -2 + 2x$

can check other basis vectors similarly.

$$\begin{aligned} (B_0^2)' = -2 + 2x &= aB_0^2 + bB_1^2 + cB_2^2 \\ &= a(1-2x+x^2) + b(2x-2x^2) + c(x^2) \\ &= 1 \cdot a + x(-2a+2b) + x^2(a-2b+c) \end{aligned}$$

Compare coefficients: $-2 = 1 \cdot a = a$

$$2 = -2(-2) + 2b \Rightarrow -2 = 2b \Rightarrow b = -1$$

$$0 = (-2) + (-2)(-1) + c \Rightarrow c = 0$$

$$\Rightarrow 1^{st} \text{ column of } M_{SS}^L = \begin{pmatrix} -2 \\ -1 \\ 0 \end{pmatrix}$$

Elementary Matrices

Elementary row transformations as used in Gaussian elimination can be viewed as left-multiplication by a matrix.

Type 1: exchange two rows of a matrix

$$P_{ij} = \begin{pmatrix} \ddots & & & & & \\ & \ddots & & & & \\ & & 0 & 1 & & \\ & & 1 & 0 & & \\ & & & & \ddots & \\ & & & & & \ddots \end{pmatrix} \begin{matrix} \\ \\ i \\ j \\ \\ \end{matrix}$$

$$\text{check: } P_{ij}x = \begin{pmatrix} x_1 \\ \vdots \\ x_{i-1} \\ x_j \\ x_{i+1} \\ \vdots \\ x_{j-1} \\ x_i \\ x_{j+1} \\ \vdots \\ x_n \end{pmatrix}$$

$$P_{ij}^{-1} = P_{ij}$$

Type 2: Multiply row i by non-zero scalar α

$$M_i(\alpha) = \begin{pmatrix} \ddots & & & & \\ & \ddots & & & \\ & & \alpha & & \\ & & & \ddots & \\ & & & & \ddots \end{pmatrix} \begin{matrix} \\ \\ i \\ \\ \end{matrix}$$

$$M_i(\alpha)x = \begin{pmatrix} x_1 \\ \vdots \\ x_{i-1} \\ \alpha x_i \\ x_{i+1} \\ \vdots \\ x_n \end{pmatrix}$$

$$M_i(\alpha)^{-1} = M_i\left(\frac{1}{\alpha}\right)$$

recall $\alpha \neq 0$

Type 3: Add the α -multiple of row j to row i

$$A_{ij}(\alpha) = \begin{pmatrix} \ddots & & & & \\ & \ddots & & & \\ & & \alpha & 1 & \\ & & & \ddots & \\ & & & & \ddots \end{pmatrix} \begin{matrix} \\ \\ j \\ i \\ \end{matrix}$$

$$A_{ij}(\alpha)x = \begin{pmatrix} x_1 \\ \vdots \\ x_{i-1} \\ x_i + \alpha x_j \\ x_{i+1} \\ \vdots \\ x_n \end{pmatrix}$$

$$A_{ij}(\alpha)^{-1} = A_{ij}(-\alpha)$$

(try for yourself)

Def.: A is in row-echelon form (REF) if

(i) the first non-zero entry in each row is strictly to the right of the first non-zero entry of the row above. These elements are called pivots.

[(ii) all non-zero rows are above zero rows]

E.g.
$$\begin{pmatrix} 1 & 2 & 3 & 4 \\ 0 & 1 & 2 & 3 \\ 0 & 0 & 5 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix} \text{ is REF}$$

$$\begin{pmatrix} 5 & 1 & 1 \\ 0 & 0 & 6 \\ 0 & 0 & 0 \end{pmatrix} \text{ is REF}$$

$$\begin{pmatrix} 2 & 3 & 1 & 5 \\ 5 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix} \text{ is not REF}$$

a matrix is in reduced REF (RREF), if, in addition

(iii) All pivots equal 1

(iv) The pivot is the only non-zero element in its column.

E.g.
$$\begin{pmatrix} 1 & 0 & 2 & 3 \\ 0 & 1 & 5 & 6 \\ 0 & 0 & 0 & 0 \end{pmatrix} \text{ is RREF}$$

$$\begin{pmatrix} 1 & 0 & 2 & 3 \\ 0 & 0 & 1 & 2 \\ 0 & 0 & 0 & 0 \end{pmatrix} \text{ is NOT RREF}$$

Fact: Any matrix can be brought into RREF by elementary row operations, i.e. by left-multiplying elementary matrices. (Gauss algorithm)

Def.: If $B = E_1 \cdots E_k A$, then "B is row-equivalent to A"

Theorem: $A \in M_{nn}(\mathbb{F})$ is invertible $\Leftrightarrow A$ is "row equivalent" to I .

Proof: " $\Rightarrow$ ": Let A be invertible.

We know there is $B = \underbrace{E_1 \cdots E_k}_{\text{elementary matrices}} A$ such that B is in RREF

$$\Rightarrow BA^{-1} = E_1 \cdots E_k$$

$$\Rightarrow BA^{-1}E_k^{-1} \cdots E_1^{-1} = I$$

Claim: $B = I$, Because: suppose not.

Since B is RREF, it must have a zero row.

But then $B \cdot *$ has a zero row, so it cannot be I $\Downarrow$

So A is row-equivalent to I

" $\Leftarrow$ ": $I = E_1 \cdots E_k A$

$$\Rightarrow \underbrace{E_k^{-1} \cdots E_1^{-1}}_{\text{invertible}} = A$$

$\Rightarrow A$ is invertible

□

and $A^{-1} = E_1 \cdots E_k$, which justifies the "naïve" procedure to invert matrices.