

Last Class:

Θ Methods:

$$\vec{u}_{j+1} = \vec{u}_j + \Theta \lambda D \vec{u}_j + (1-\Theta) \lambda D \vec{u}_{j+1} + \vec{f}_j$$

$\Theta = 0$ B.E

$\Theta = 1$ Explicit

Applied VN analysis which suggested stability if

$$(2\Theta - 1) \lambda \leq \frac{1}{2} \quad B$$

Discrete Max principle

$$\lambda \Theta \leq \frac{1}{2} \quad A$$

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Applied VN analysis which suggested stability if

$$(2\Theta - 1) \lambda \tau \leq \frac{1}{2}$$

Discrete Max principle

$$\lambda \Theta \leq \frac{1}{2}$$

$$\lambda \leq \frac{1}{2}$$

$$k \leq \frac{1}{2} h^2$$

$$O(k^2) + O(h^2)$$

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$$\vec{u}_{j+1} = \vec{u}_j + \Theta \lambda D \vec{u}_j + (1-\Theta) \lambda D \vec{u}_{j+1} + \vec{f}_j$$

$\Theta = 0$ B.E

$\Theta = 1$ Explicit

always stable
 $\forall 0 \leq \Theta \leq \frac{1}{2}$

Applied VN analysis which suggested stability if

$$(2\Theta - 1)\lambda \leq \frac{1}{2} \quad B$$

Discrete Max principle

$$\lambda \Theta \leq \frac{1}{2} \quad A$$

In fact: $\textcircled{A} \Rightarrow \textcircled{B}$

$$\lambda\theta < \frac{1}{2}$$

$$(2\theta - 1)\lambda =$$

In fact: $\textcircled{A} \Rightarrow \textcircled{B}$

$$(2\theta - 1)\lambda = \theta\lambda + \sqrt{(\theta-1)\lambda} \leq 0$$

$0 \leq \theta \leq 1$
 $-1 \leq \theta - 1 \leq 0$

$$\leq \theta\lambda \leq \frac{1}{2}$$

Convergence under strong condition A $\theta\lambda \leq \frac{1}{2}$

$U_{i,j}$

$$\begin{aligned} U_{i,j+1} (1 + 2(1-\theta)\lambda) &= (1-\theta)\lambda (U_{i-1,j+1} + U_{i+1,j+1}) \\ &\quad + \lambda\theta (U_{i-1,j} + U_{i+1,j}) \\ &\quad + (1-2\theta\lambda) U_{i,j} + k^f_{i,j} \end{aligned}$$

Convergence under strong condition A

$U_{i,j}$

$$U_{i,j+1} (1 + 2(1-\theta)\lambda) = (1-\theta)\lambda (U_{i-1,j+1} + U_{i+1,j+1}) \\ + \lambda\theta (U_{i-1,j} + U_{i+1,j}) \\ + (1-2\theta\lambda) U_{i,j} + k f_{i,j}$$

$$u_{i,j+1} (1 + 2(1-\theta)\lambda) = (1-\theta)\lambda (u_{i-1,j+1} + u_{i+1,j+1}) \\ + \lambda\theta (u_{i-1,j} + u_{i+1,j}) \\ + (1-2\theta\lambda) u_{i,j} + k f_{i,j} + k \tilde{c}_{i,j}$$

$$\theta \lambda \leq \frac{t}{2} \Rightarrow 1 - 2\theta \lambda \geq 0$$

$$\begin{aligned} E_{i,j+1} \cdot (1 + 2(1-\theta)\lambda) &= (1-\theta)\lambda (E_{i+1,j+1} + E_{i-1,j+1}) \\ &\quad + \theta\lambda (E_{i+1,j} + E_{i-1,j}) \\ &\quad + (1-2\theta)\lambda E_{i,j} - k \tau_{i,j} \\ &\geq 0 \end{aligned}$$

$$\begin{aligned}
 E_{i,j+1} \cdot (1 + 2(1-\theta)\lambda) &= (1-\theta)\lambda (E_{i+1,jH} + E_{i-1,jH}) \\
 &\quad + \theta\lambda (E_{i+1,j} + E_{i-1,j}) \\
 &\quad + (1-2\theta)\lambda E_{i,j} - k \tilde{\epsilon}_{i,j}
 \end{aligned}$$

$$E_j = \max_i |E_{i,j}| \quad (\|\vec{E}_j\|_\infty)$$

$$\begin{aligned}
 E_{j+1} (1 + 2(1-\theta)\lambda) &\leq 2(1-\theta)\lambda E_{jH} + 2\theta\lambda E_j \\
 &\quad + \underbrace{(1-2\theta)\lambda E_j + k \tilde{\epsilon}_j}_{\hookrightarrow \text{used } \theta\lambda \leq \frac{1}{2}}
 \end{aligned}$$

$$\tilde{\epsilon}_j = \max_i |\tilde{\epsilon}_{i,j}|$$

$$E_0 \leq \max_j E_j$$

$$E_{j+1} \leq E_j + k z_j$$

Exercise:

$$\max_j E_j \leq E_0 + Mk \max_j z_j$$

$$= E_0 + T C$$

$$E_0 \leq E_0 + \bar{T} \bar{C}$$

$\bar{C} = \max_{i,j} |C_{i,j}|$

$$\max_j \max_i |E_{i,j}| = \max_{i,j} |V_{i,j} - u_{i,j}| \quad E_0 = 0$$

$$\max_{i,j} |V_{i,j} - u_{i,j}| \leq T \max_{i,j} |C_{i,j}|$$

$$E_{j+1} \leq E_j + k \bar{z}_j$$

Exercise:

$$\max_j E_j \leq E_0 + M k \max_j \bar{z}_j$$
$$= E_0 + T \bar{c}$$

$$\bar{c} = \max_{i,j} |\bar{c}_{i,j}|$$

I.e.

$$\max_{i,j} |U_{i,j} - u_{i,j}| \leq T \max_{i,j} |\bar{c}_{i,j}|$$

$$\theta \lambda \leq \frac{1}{2}$$

$$E_{j+1} \leq E_j + k \bar{\epsilon}_j$$

$$k \leq \frac{1}{2\theta} h^2$$

Exercise:

$$\max_j E_j \leq E_0 + Mk \max_j \bar{\epsilon}_j$$
$$= E_0 + T \bar{\epsilon}$$

$$\bar{\epsilon} = \max_{i,j} |\bar{\epsilon}_{i,j}|$$

I.e.

$$\max_{i,j} |U_{i,j} - u_{i,j}| \leq T \max_{i,j} |\bar{\epsilon}_{i,j}|$$

So $\bar{\epsilon}_{i,j} = O(k) + O(h^2) \Rightarrow \max_{i,j} |U_{i,j} - u_{i,j}| \rightarrow 0$

Next: convergence using only $(2\theta - 1)x \leq \frac{1}{2}$ B

under a different notion of error.

$$\vec{U}_j \quad U_{i,j} \quad \vec{E}_j = \vec{U}_j - \vec{u}_j \leftarrow N\text{-dimensional vector}$$

$\vec{u}_j \quad u_{i,j}$ we've been working with

$$\|\vec{E}_j\|_\infty = \max_i |E_{i,j}|$$

Other vector norms:

$$1 \leq p < \infty$$

$$\|x\|_p = \left[\sum_{i=1}^n |x_i|^p \right]^{1/p}$$

$$x \in \mathbb{R}^n$$
$$x = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$$

Other vector norms:

$$1 \leq p < \infty$$

$$\|x\|_p = \left[\sum_{i=1}^n |x_i|^p \right]^{1/p}$$

$$p=1 \quad \sum_{i=1}^{\infty} |x_i|$$

Other vector norms:

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$$\|x\|_p = \left[\sum_{i=1}^n |x_i|^p \right]^{1/p}$$

$$p=1 \quad \sum_{i=1}^{\infty} |x_i|$$

$$\|x\|_\infty = \max |x_i|$$

$$p=2 \quad \left[\sum_{i=1}^{\infty} |x_i|^2 \right]^{1/2}$$

Other vector norms:

$$0 \leq \frac{1}{p} \leq 1$$

$$1 \leq p < \infty$$

$$\|x\|_p = \left[\sum_{i=1}^n |x_i|^p \right]^{1/p}$$

$\|x\|_p$
norm(x ,
 p)
 l_1, l_2, l_∞

$$p=1$$

$$\sum_{i=1}^{\infty} |x_i|$$

$$\|x\|_\infty = \max |x_i|$$

$$p=2$$

$$\left[\sum_{i=1}^{\infty} |x_i|^2 \right]^{1/2}$$

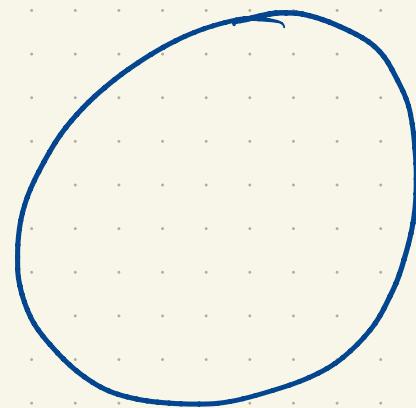
Exercise: $\|x\|_p \rightarrow \|x\|_\infty$

as $p \rightarrow \infty$

Unit Balls : $\{x : \|x\|_p \leq 1\} = B_p(r)$

$$\|x\|_2 \leq 1$$

Unit Balls : $\{x : \|x\|_p \leq 1\} = B_p(r)$



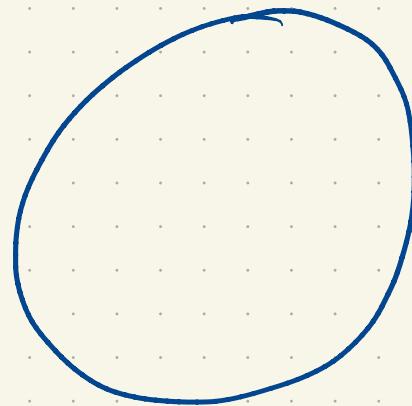
$$\|x\|_2 = 1$$

Unit Balls: $\{x : \|x\|_p \leq 1\} = B_p(r)$

$$\bar{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$$\|x\|_\infty$$

$$\max(|x_1|, |x_2|)$$

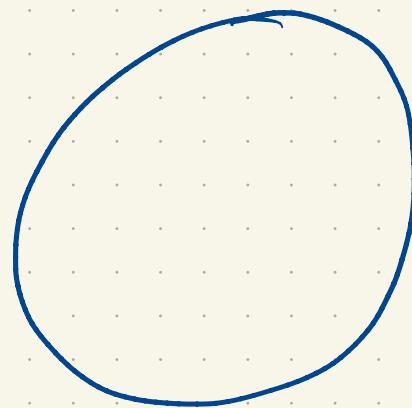


$$\|x\|_2 = 1$$

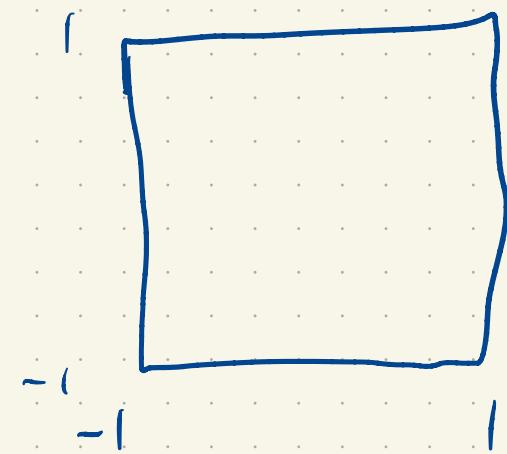
$$\|x\|_\infty = 1$$

$|x_1| = 1, |x_2| \leq 1$

Unit Balls: $\{x : \|x\|_p \leq 1\} = B_p(r)$



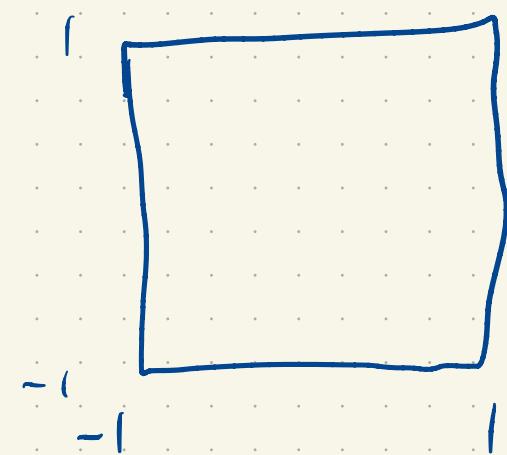
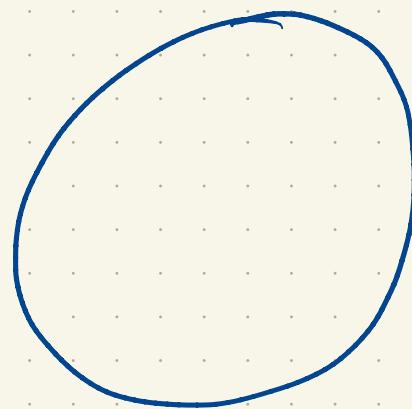
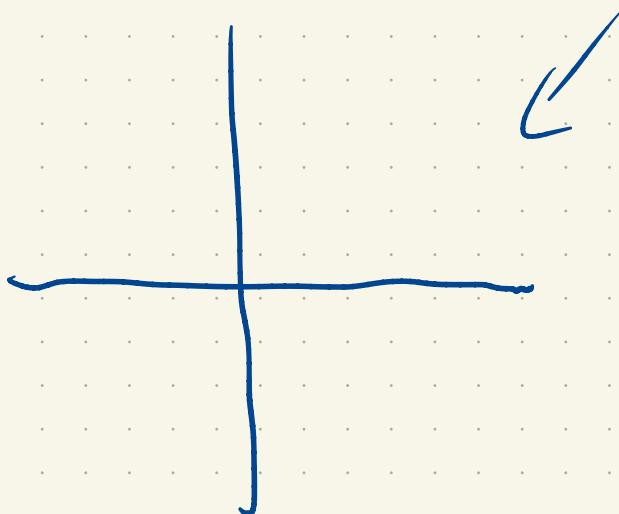
$$\|x\|_2 = 1$$



$$\|x\|_\infty = 1$$

Unit Balls: $\{x : \|x\|_p \leq 1\} = B_p(r)$

$$|x_1| + |x_2| = 1 \rightarrow x_1 + x_2 = 1$$

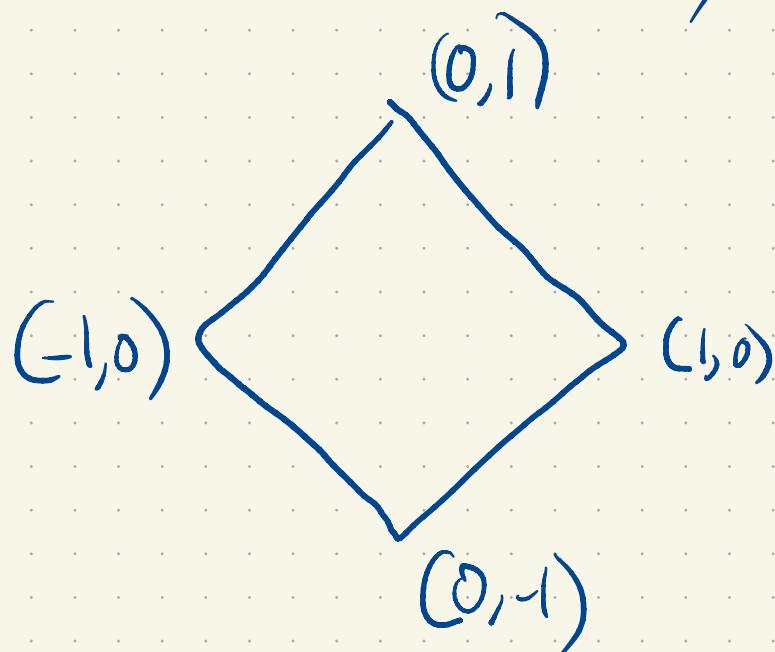
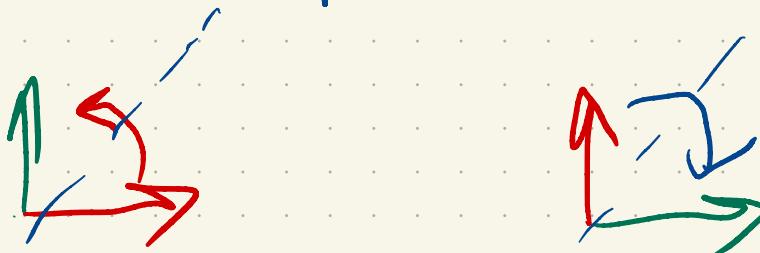


$$\|x\|_1 = 1$$

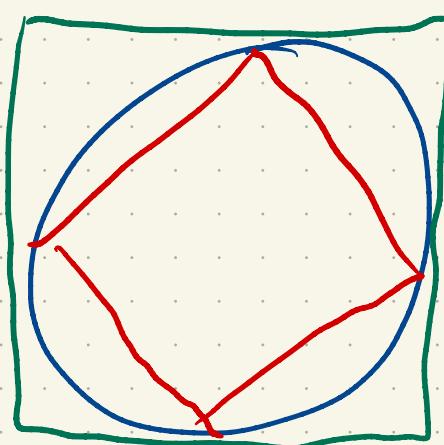
$$\|x\|_2 = 1$$

$$\|x\|_\infty = 1$$

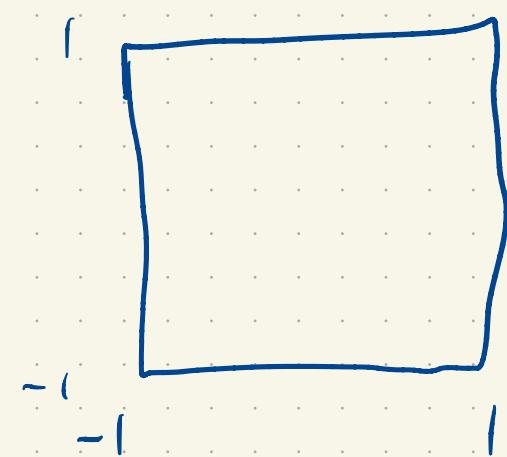
Unit Balls: $\{x : \|x\|_p \leq 1\} = B_p(r)$



$$\|x\|_1 = 1$$



$$\|x\|_2 = 1$$



$$\|x\|_\infty = 1$$

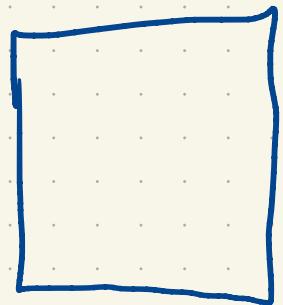
$$\|A + B\|_p \leq \|A\|_p + \|B\|_p$$

$$\det(A+B)$$

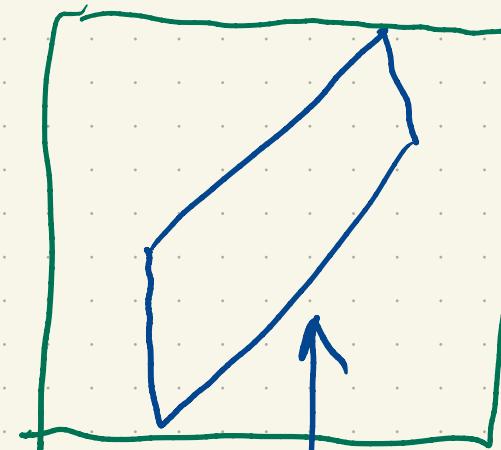
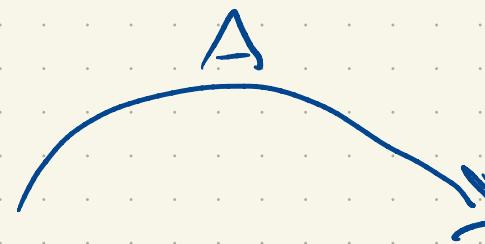
$$\det(AB) = \det(A)\det(B)$$

$$\det: GL(\mathbb{R}^n) \rightarrow \mathbb{R}$$

Matrix Norms A : $n \times n$ matrix



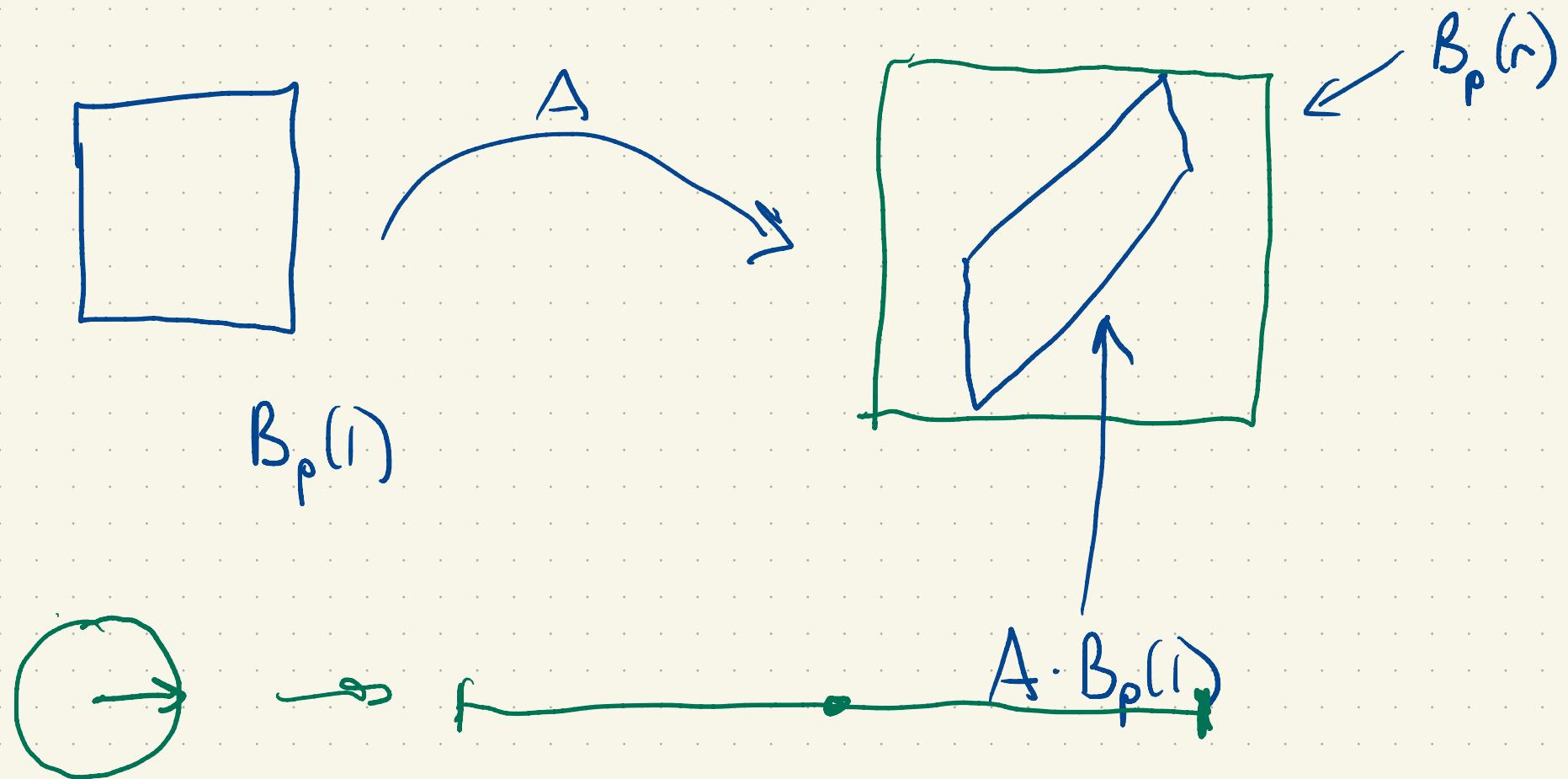
$$B_p(1)$$



$$B_p(r)$$

$$A \cdot B_p(1)$$

Matrix Norms A : $n \times n$ matrix



$$\|A\|_p \text{ is smallest } r \text{ so } A B_p(1) \subseteq B_p(r)$$