

# Gaussian Elimination

3 operations

- ① Multiplying a row by a nonzero scalar
- ② Swapping rows
- ③ Adding a scalar multiple of a row to another row

Reduced row echelon form

$$\left[ \begin{array}{cccc|c} 1 & 0 & * & 0 & * \\ 0 & 1 & * & 0 & * \\ 0 & 0 & 1 & 0 & * \\ 0 & 0 & 0 & 0 & 0 \end{array} \right]$$

Linear Equations

- ① Superposition: If  $x+y = z \rightarrow f(y+z) = f(y) + f(z)$
- ② Homogeneity:  $f(\alpha x) = \alpha f(x)$

$$f(x) = b^2 x \quad \left. \begin{array}{l} f(\alpha x) = b^2 \alpha x \\ \alpha f(x) = b^2 \alpha x \end{array} \right\} b^2 \alpha x = b^2 \alpha x \quad \checkmark \quad \text{homogeneity}$$

$$\left. \begin{array}{l} f(y+z) = b^2(y+z) \\ f(y) + f(z) = b^2 y + b^2 z \end{array} \right\} = \quad \checkmark \quad \text{superposition}$$

## Vectors / Matrices

$$\begin{aligned} \vec{x} + \vec{y} &= \vec{y} + \vec{x} \\ (\vec{x} + \vec{y}) + \vec{z} &= \vec{x} + (\vec{y} + \vec{z}) \\ \vec{x} + \vec{0} &= \vec{x} \\ \vec{x} + (-\vec{x}) &= \vec{0} \end{aligned} \quad \boxed{\text{Addition}}$$

$$\begin{aligned} (\alpha\beta)\vec{x} &= \alpha(\beta\vec{x}) \\ (\alpha + \beta)\vec{x} &= \alpha\vec{x} + \beta\vec{x} \\ 1\vec{x} &= \vec{x} \end{aligned} \quad \boxed{\text{scalar multiplication}}$$

$$\vec{x} = \begin{bmatrix} x^1 \\ \vdots \\ x_n \end{bmatrix} \quad \text{then} \quad \vec{x}^T = [x^1 \dots x_n]$$

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

Linear Transformation:  $f_A(\vec{x}) = A\vec{x}$

$$\begin{aligned} \text{ccw} &\left[ \begin{array}{cc} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{array} \right] \\ \text{cw} &\left[ \begin{array}{cc} \cos \theta & \sin \theta \\ -\sin \theta & \cos \theta \end{array} \right] \end{aligned} \quad \boxed{\text{rotation}}$$

$$\vec{y}^T \vec{x} = [y_1 \ y_2 \ \dots \ y_n] \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = y_1 x_1 + y_2 x_2 + \dots + y_n x_n$$

$$\begin{array}{c} \text{Row 1:} \\ \boxed{\begin{bmatrix} A_{11} & A_{12} & \dots & A_{1n} \\ A_{21} & A_{22} & \dots & A_{2n} \\ \vdots & \vdots & & \vdots \\ A_{m1} & A_{m2} & \dots & A_{mn} \end{bmatrix}} \times \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} A_{11}x_1 + A_{12}x_2 + \dots + A_{1n}x_n \\ A_{21}x_1 + A_{22}x_2 + \dots + A_{2n}x_n \\ \vdots \\ A_{m1}x_1 + A_{m2}x_2 + \dots + A_{mn}x_n \end{bmatrix} \end{array}$$

$$\begin{array}{c} \text{Row 2:} \\ \boxed{\begin{bmatrix} A_{11} & A_{12} & \dots & A_{1n} \\ \cancel{A_{21}} & \cancel{A_{22}} & \dots & \cancel{A_{2n}} \\ \vdots & \vdots & & \vdots \\ A_{m1} & A_{m2} & \dots & A_{mn} \end{bmatrix}} \times \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} A_{11}x_1 + A_{12}x_2 + \dots + A_{1n}x_n \\ \cancel{A_{21}}x_1 + \cancel{A_{22}}x_2 + \dots + \cancel{A_{2n}}x_n \\ \vdots \\ A_{m1}x_1 + A_{m2}x_2 + \dots + A_{mn}x_n \end{bmatrix} \end{array}$$

$$\begin{array}{c} \vdots \\ \text{Row } m: \\ \boxed{\begin{bmatrix} A_{11} & A_{12} & \dots & A_{1n} \\ A_{21} & A_{22} & \dots & A_{2n} \\ \vdots & \vdots & & \vdots \\ A_{m1} & A_{m2} & \dots & A_{mn} \end{bmatrix}} \times \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} A_{11}x_1 + A_{12}x_2 + \dots + A_{1n}x_n \\ A_{21}x_1 + A_{22}x_2 + \dots + A_{2n}x_n \\ \vdots \\ A_{m1}x_1 + A_{m2}x_2 + \dots + A_{mn}x_n \end{bmatrix} \end{array}$$

$$\text{x-axis} \quad \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix} \quad \text{y-axis} \quad \begin{bmatrix} -1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$y=x \quad \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \quad y=-x \quad \begin{bmatrix} 0 & -1 \\ -1 & 0 \end{bmatrix}$$

Reflect (R) then rotate (O)  
 $R\vec{v}$  then multiply w/ O  $\rightarrow O(R\vec{v})$

## Linear (in)dependence

Linear dependence: If vector can be written as combo of other vectors  
 $a_1\vec{v}_1 + \dots + a_n\vec{v}_n = \vec{0}$  and not all  $a_i$ 's = 0

Independence: if  $a_1\vec{v}_1 + \dots + a_n\vec{v}_n = \vec{0}$  implies  $a_1 = \dots = a_n = 0$

Span: set of all linear combinations of  $\{\vec{v}_1 \dots \vec{v}_n\}$   
 $\hookrightarrow$  range/column space  
 \* can solve w/ Gaussian Elimination

$$\begin{bmatrix} x_1 & y_1 \\ x_2 & y_2 \end{bmatrix} \begin{bmatrix} a_1 \\ a_2 \end{bmatrix} = \begin{bmatrix} 6 \\ 0 \end{bmatrix}$$

$$\text{span} \left\{ \sum_{i=1}^n a_i \vec{v}_i \mid a_i \in \mathbb{R} \right\}$$

# Proofs

\* review proof examples

- ① Write down what you know (rephrasing or in math)
- ② Write down what you want to show (map out your path)
- ③ Find similarities (how can I form look like the other)
- ④ Try a simple example for intuition
- ⑤ Manipulate both sides of claim & JUSTIFY each step

## State Transition Matrices / Inverses

$$\begin{bmatrix} P_{A \rightarrow A} & P_{A \rightarrow B} & P_{A \rightarrow C} \\ P_{B \rightarrow A} & P_{B \rightarrow B} & P_{B \rightarrow C} \\ P_{C \rightarrow A} & P_{C \rightarrow B} & P_{C \rightarrow C} \end{bmatrix} \quad \begin{array}{l} \text{Given current state: } \vec{v}[t] \\ \text{state transition matrix: } A \\ \vec{v}[t+1] = A\vec{v}[t] \end{array}$$

A square matrix  $M$  and its inverse  $M^{-1}$  satisfies  $MM^{-1} = I$

$$\text{Let } M = \begin{bmatrix} 1 & 1 \\ 2 & 1 \end{bmatrix} \text{ and } M^{-1} = \begin{bmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{bmatrix}$$

$$\hookrightarrow \begin{bmatrix} 1 & 1 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \quad [M | I_n] \rightarrow [I_n | M^{-1}]$$

$$\hookrightarrow \left[ \begin{array}{cc|cc} 1 & 1 & 1 & 0 \\ 2 & 1 & 0 & 1 \end{array} \right] \quad * \text{ use Gaussian Elimination} \quad M^{-1} = \frac{1}{ad-bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$$

$A$  is invertible  $\iff$  equation  $A\vec{x} = \vec{b}$  has a unique solution

$A$  is invertible  $\iff A$  has linearly independent columns

$$AB = BA = I$$

## Vector Spaces

\* review problem solving techniques

vector space  $V$  is a set of vectors that satisfies  
vector addition  
see properties  $\rightarrow$  scalar multiplication

Basis = series of vectors that defines a vector space  $\{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_n\}$

- ①  $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_n$  must be linearly independent
- ② For any vector  $\vec{v} \in V$ ,  $\vec{v} = a_1\vec{v}_1 + a_2\vec{v}_2 + \dots + a_n\vec{v}_n$  (scalars)
- \* minimum set of vectors needed to represent vector space

Dimension = # of basis vectors

Subspace  $U$  = subset of vector space  $V$

- ① contains the 0 vector:  $\vec{0} \in U$
- ② closed under vector addition:  $\vec{v}_1, \vec{v}_2 \in U \rightarrow \vec{v}_1 + \vec{v}_2$  must be in  $U$
- ③ closed under scalar multiplication:  $\vec{v} \in U$  & scalar  $\alpha \in \mathbb{R}$ ,  $\alpha\vec{v}$  in  $U$

$$\text{Col}(A) = \left\{ \vec{v} \mid \vec{v} = \sum_{i=1}^m x_i \vec{a}_i \text{ where } x_i \text{'s are scalars} \right.$$

\* To solve  $\rightarrow$  use Gaussian elimination and columns w/  
pivot = vectors in the span

$$\begin{aligned} D &= \begin{bmatrix} 1 & -1 & -3 & 4 \\ 3 & -3 & -5 & 8 \\ 1 & -1 & -1 & 2 \end{bmatrix} & \xrightarrow{\text{Row operations}} & \begin{bmatrix} 1 & -1 & -3 & 4 \\ 0 & 0 & 1 & -1 \\ 0 & 0 & 1 & -1 \end{bmatrix} \\ (1) - 3(1) &\left[ \begin{array}{cccc} 1 & -1 & -3 & 4 \\ 0 & 0 & 1 & -4 \\ 1 & -1 & -1 & 2 \end{array} \right] & \xrightarrow{\text{Row operations}} & \left[ \begin{array}{cccc} 1 & -1 & -3 & 4 \\ 0 & 0 & 1 & -1 \\ 0 & 0 & 0 & 0 \end{array} \right] \\ (3) - (1) &\left[ \begin{array}{cccc} 1 & -1 & -3 & 4 \\ 0 & 0 & 1 & -4 \\ 0 & 0 & 2 & -2 \end{array} \right] & \xrightarrow{\text{Row operations}} & \text{Col}(D) = \text{span} \left\{ \begin{bmatrix} 1 \\ 3 \\ 1 \end{bmatrix}, \begin{bmatrix} -3 \\ -5 \\ -1 \end{bmatrix} \right\} \text{ dim}=2 \end{aligned}$$

$\text{RANK}(A) = \dim(\text{COL}(A)) \leq \min(m, n)$  (# of linearly independent cols)

Nullspace: set of vectors mapped to 0 by A  $\rightarrow \{\vec{x} \mid A\vec{x} = \vec{0}, \vec{x} \in \mathbb{R}^n\}$   $A\vec{x} = \vec{0}$

\* use GA to set matrix = 0  $\rightarrow$  create an  $\vec{x}$  for  $x_1, \dots, x_n$  in matrix  
→ scalar multiplication

$$\left[ \begin{array}{ccccc|c} 1 & -1 & -3 & 4 & | & 0 \\ 0 & 0 & 1 & -1 & | & 0 \\ 0 & 0 & 0 & 0 & | & 0 \end{array} \right] \xrightarrow{(1)+3(2)} \left[ \begin{array}{ccccc|c} 1 & -1 & 0 & 1 & | & 0 \\ 0 & 0 & 1 & -1 & | & 0 \\ 0 & 0 & 0 & 0 & | & 0 \end{array} \right] \quad \begin{matrix} x_1 = x_2 - x_3 \\ x_3 = x_4 \\ x_4 = \alpha \end{matrix} \quad \begin{matrix} \text{let } x_2 = \beta \\ x_4 = \beta \end{matrix}$$

$$\vec{x} = \begin{bmatrix} \alpha - \beta \\ \beta \\ \beta \\ \beta \end{bmatrix} = \alpha \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} + \beta \begin{bmatrix} -1 \\ 1 \\ 0 \\ 1 \end{bmatrix}$$

$$\text{NULL}(0) = \text{span} \left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix} \right\} \quad \dim = 2$$

$\text{COL}(A) = \# \text{ of linearly independent cols}$

$\text{NULL}(A) = \# \text{ of linearly dependent cols}$

$$m - \dim(\text{COL}(A)) = \dim(\text{N}(A))$$

Rank-nullity theorem

NON-TRIVIAL: columns = linearly dependent  
TRIVIAL: INDEPENDENT  $\rightarrow \vec{0}$

## Eigenvector / Eigenvalues

Eigenvector: nonzero vector  $A\vec{x} = \lambda\vec{x}$  where  $\lambda$  is eigenvalue of  $\vec{x}$

$$A\vec{x} = \lambda\vec{x} = \vec{0}$$

$$(A - \lambda I_n)\vec{x} = \vec{0} \rightarrow \det(A - \lambda I) = 0$$

\* All eigenvectors w/ diff eigenvalues are linearly indep

$$\underbrace{\lambda^2 - (a+d)\lambda + (ad-bc)}_0 = 0$$

2 distinct real eigenvals → linearly independent eigenvectors

characteristic polynomial

Steady state freq → eigenvector associated with  $\lambda=1$  and normalize so that the columns sum to 1

\* review last dls

$$\vec{x}[t] = \alpha_1(\lambda_1^t \vec{v}_1) + \alpha_2(\lambda_2^t \vec{v}_2) + \dots + \alpha_n(\lambda_n^t \vec{v}_n)$$

\* want to know if  $\vec{x}[t]$  will converge

- ① If  $|\lambda_i| > 1$  then  $\lambda_i^t \vec{v}_i \rightarrow \infty$
  - ② If  $\lambda_i = -1$  then  $\lambda_i^t \vec{v}_i$  will oscillate
  - ③ If all  $\lambda_i$   $-1 < \lambda_i \leq 1$  then each term  $\rightarrow 0$  ( $\lambda_i \neq 1$ ) or stay the same ( $\lambda_i = 1$ )  
 $\hookrightarrow \vec{x}[t]$  will always converge to a fixed value
- } fail to converge

## Properties

For a square matrix A:

- ① A is invertible
- ②  $A\vec{x} = \vec{b}$  has a unique soln for any  $\vec{b}$
- ③ A has linearly independent cols
- ④ A has a trivial null space
- ⑤ Determinant of A  $\neq 0$

$$|A \cdot B| = |A| \cdot |B| \quad \lambda \text{ of } A \neq 0$$

Matrix & transpose have same determinants

$$\lim_{n \rightarrow \infty} \vec{x}[n] = A^n \vec{x}[0]$$

$$\begin{aligned} &= A^n [\alpha_1 \vec{v}_1 + \alpha_2 \vec{v}_2 + \dots + \alpha_n \vec{v}_n] \\ &= \alpha_1 (A^n \vec{v}_1) + \alpha_2 (A^n \vec{v}_2) + \dots + \alpha_n (A^n \vec{v}_n) \\ &= \alpha_1 (\lambda_1^n \vec{v}_1) + \alpha_2 (\lambda_2^n \vec{v}_2) + \dots \end{aligned}$$

\* If invertible matrix A has eigenvalue  $\lambda \rightarrow$  then  $A^{-1}$  has eigenvalue  $\frac{1}{\lambda}$

\* All vectors in the nullspace of a matrix are in its eigenspace for  $\lambda = 0$   $A\vec{x} = \vec{0} \rightarrow A\vec{x} = 0\vec{x}$

\* REVIEW HW PROOFS

\* even if  $\lambda > 1 \rightarrow$  doesn't necessarily mean it'll diverge

\* LOOK at equivalent definitions + expand stuff out

$G = M_1 M_2$  (After know  $M_1$  and  $M_2$  have inverses bc linear indep)

$$M_1^{-1} G = M_1^{-1} M_1 M_2$$

$$M_1^{-1} G = M_2$$

$$M_2^{-1} M_1^{-1} G = M_2^{-1} M_2$$

$$\underline{M_2^{-1} M_1^{-1} G = I}$$

$$\underline{G^{-1}}$$

$$A B = I$$

$$\text{Then } B = A^{-1}$$

$$A = B^{-1}$$

$$\begin{array}{r} \cancel{14} \\ -70 \\ \hline \end{array} \quad \begin{array}{r} \cancel{14} \\ \times \cancel{14} \\ \hline 56 \\ \hline \end{array} \quad \begin{array}{r} \cancel{56} \\ -25 \\ \hline 45 \\ \hline \end{array} \quad \begin{array}{r} \cancel{146} \\ -45 \\ \hline 151 \\ \hline \end{array}$$

$$\lambda_1 = \frac{\xi}{2} \quad (a - \frac{\xi}{2})(a - \frac{\xi}{2}) - b^2 = 0 \quad a^2 - 5a + \frac{2\xi}{4} - b^2 = 0$$

$$\left(\frac{14}{4}\right)^2 - \frac{70}{4} + \frac{25}{4} - b^2 = 0$$

$$\lambda_2 = \frac{\eta}{2} \quad (a - \frac{\eta}{2})(a - \frac{\eta}{2}) - b^2 = 0 \quad a^2 - 9a + \frac{81}{4} - b^2 = 0$$

$$\frac{146}{4} - \frac{45}{4} = b^2$$

$$\sqrt{\frac{151}{4}} = \sqrt{b^2}$$

$$\cancel{a^2} - 5a + \frac{25}{4} - \cancel{b^2} = \cancel{a^2} - 9a + \frac{81}{4} - \cancel{b^2}$$

$$\frac{781}{-25} \quad 4a = \frac{81}{4} - \frac{25}{4}$$

$$\frac{-25}{56} \quad 4a = \frac{56}{4}$$

$$4a = 14$$

$$a = 14/4$$

# KVL & KCL & Circuit Elements

Wire:  $V = 0$

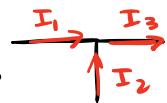
Resistor:  $V = IR$

Open circuit:  $I = 0$

Voltage source:  $V = V_s$

Current source:  $I = I_s$

**KCL:** Current flowing into node = current flowing out of node :  $I_1 + I_2 = I_3$



**KVL:**  $\sum_{\text{loop}} V_k = 0$  \* if  $+ \rightarrow -$ , subtract voltage  
\* start from pos end  $- \rightarrow +$ , add voltage

Ohm's law:  $V_{\text{elem}} = I_{\text{elem}} R$

$$\text{current: } I = \frac{dQ}{dt}$$

$$\text{power: } P = IV = \frac{V^2}{R} = I^2 R$$

$$\text{resistivity: } R = \rho \times \frac{L}{A}$$

## Circuit Analysis / NVA

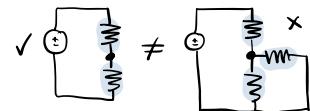
- Passive sign convention : enter positive exit negative

\* node = region of circuit w/ same voltage throughout

- ① Pick reference node  $\rightarrow$  label 0V
- ② Label other nodes
- ③ Label currents
- ④ Add  $+/ -$  labels on non-wire elements
- ⑤ Use KCL to write eqs at labeled nodes
- ⑥ Write I-V relationship
- ⑦ Solve w/ substitution

### Voltage Divider

\* Make sure currents equal through resistors

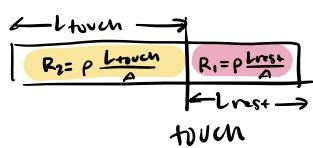


$$V_{\text{mid}} = \frac{R_2}{R_1 + R_2} V_s$$

$$\text{parallel: } R_{\text{parallel}} = \frac{R_1 R_2}{R_1 + R_2}$$

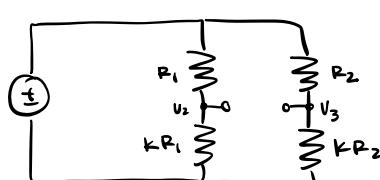
$$\text{series: } R_{\text{series}} = R_1 + R_2$$

## Resistive Touchscreens



$$V_{\text{mid}} = \frac{L_{\text{touch}}}{L} V_s$$

$P = IV$   $+P \rightarrow$  power dissipated  
 $-P \rightarrow$  power delivered

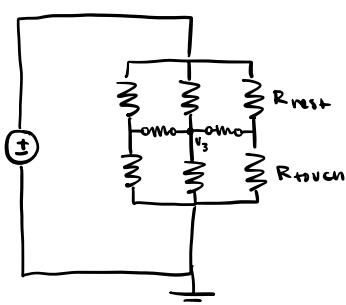


2 voltage dividers

$$U_2 = U_3 = \frac{K}{1+K} V_s$$

\* Resistor proportions equal, so same voltage drop

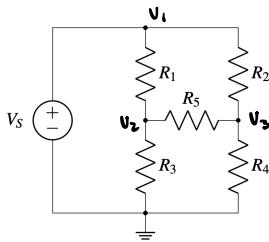
2D touchscreen



$R_{\text{rest}} = R_{\text{touch}}$  so  $U_2 = U_3 = U_4 \rightarrow$  replace horizontal resistors w/ open circuit

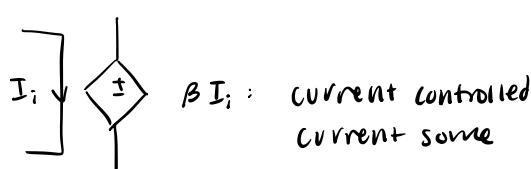
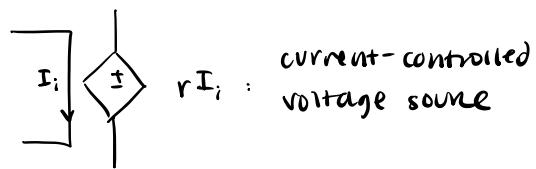
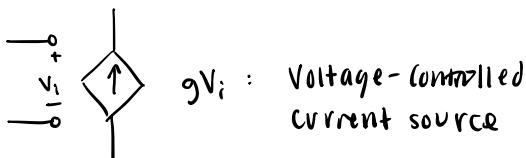
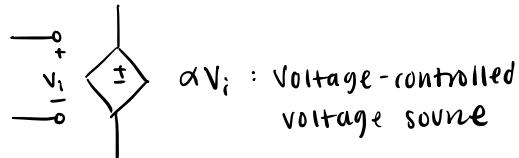
$$U_3 = \frac{L_{\text{touch}}}{L} \times V_s \quad \text{where } L_{\text{touch}} = L_{\text{touch, vertical}}$$

$$U_3 = \frac{R_{\text{touch}}}{R_{\text{touch}} + R_{\text{rest}}} \times V_s \quad \text{where } R_{\text{touch}} = \rho \frac{L_{\text{touch, horizontal}}}{A}$$



- ①  $V_1 = V_s$
- ② KCL @ unknown nodes  $K$  current flowing out of node  
 $\frac{V_2 - V_s}{R_1} + \frac{V_2 - V_3}{R_5} + \frac{V_2}{R_3} = 0$        $\frac{V_3 - V_s}{R_2} + \frac{V_3}{R_4} + \frac{V_3 - V_2}{R_5} = 0$

## Superposition + Equivalence



- ① Null one source
- ② Write KCL eqs
- ③ Use N relations & NVA
- ④ Combine final results

### SUPERPOSITION

- For each indep source

Set other sources = 0

- V source: replace w/ wire

- C source: replace w/ open circuit

Compute voltages + currents

$V_{out} = \text{sum of } V_{out,k} \text{ for all } k$

→ 2 circuits are equivalent if they have same I-V relationship

### THEVENIN & NORTON EQUIVALENT

(A) Find thevenin voltage

① ID all nodes in circuit

②  $V_{th} = V_A - V_B$

\* use nodal analysis or superposition

(B) Find norton current

① Connect a wire through A & B (short circuit)

② Simplify circuit by removing resistors (path of least resistance)

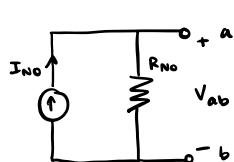
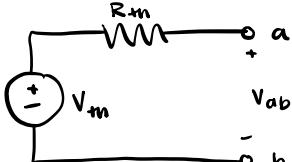
③  $I_N = I_{AB}$

(C) Find Thevenin/Norton Req

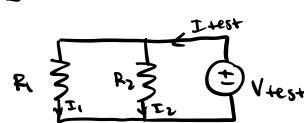
① Turn off all independent sources in circuit

② Apply test voltage  $V_{test}$  → calculate  $I_{test}$  that flows through test voltage source

$$R_{eq} = \frac{V_{test}}{I_{test}}$$



→ example



$$\begin{aligned} I_1 &= \frac{V_{test}}{R_1} \\ I_{test} &= I_1 + I_2 \\ &= \frac{V_{test}}{R_1} + \frac{V_{test}}{R_2} \end{aligned}$$

# Capacitors

\* current only if voltage changing with time

$$I = C \frac{dV_c}{dt}$$

$$\int_0^t I dt = C \int_0^t dV_c$$

$$V_c(t) = \frac{I}{C} t + V_c(0)$$

\* stores charge

$$Q = CV_c$$

$$I = C \frac{dV_c}{dt}$$

$$V_c(t) = \frac{I}{C} t + V_c(0)$$

Parallel:  $C_{eq} = C_1 + C_2$

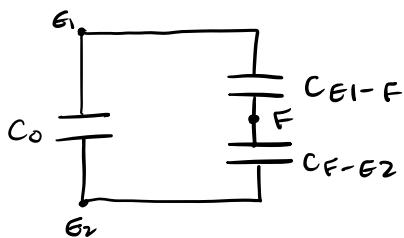
Series:  $C_{eq} = C_1 || C_2 = \frac{C_1 C_2}{C_1 + C_2}$

$$C = \epsilon \frac{A}{d}$$



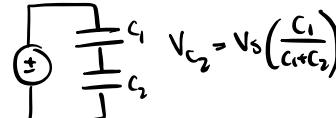
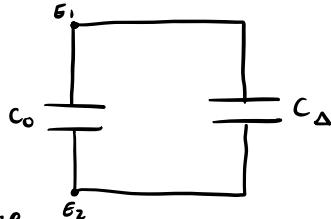
$$E = \frac{1}{2} CV^2$$

Capacitor w/ touch

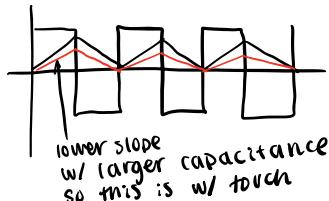


\* STEADY STATE = NO CURRENT

combine  $C_{E1-F}$  and  $C_{F-E2}$  to  $C_\Delta$



\* Apply periodic current source

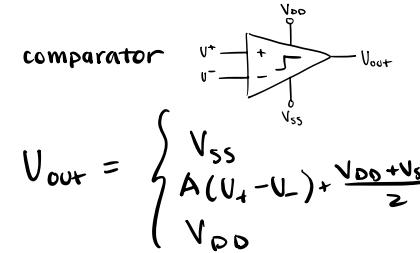


$$V_c(t) = \begin{cases} \frac{I_1}{C} t & 0 \leq t \leq \frac{T}{2} \\ -\frac{I_1}{C}(t - \frac{T}{2}) + \frac{I_1 T}{2C} & \frac{T}{2} < t \leq T \end{cases}$$

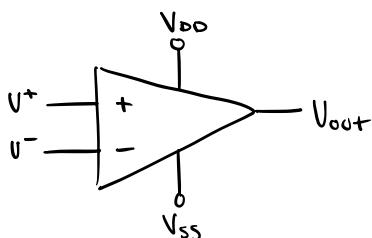
$$V_c(t) = \frac{I}{C} t + V_c(0)$$

## Op-Amps

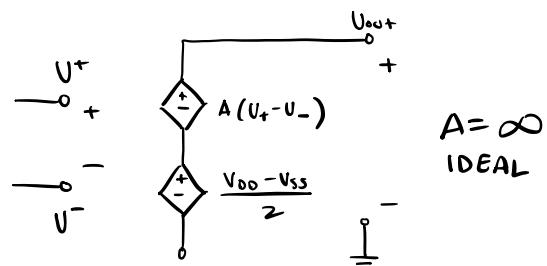
comparator



$$U_{out} = \begin{cases} V_{SS} & U^+ < U^- \\ A(U_+ - U_-) + \frac{V_{DD} + V_{SS}}{2} & U^+ > U^- \\ V_{DD} & \end{cases}$$



$$\begin{aligned} A(U_+ - U_-) + \frac{V_{DD} + V_{SS}}{2} &< V_{SS} \\ V_{SS} &\leq A(U_+ - U_-) + \frac{V_{DD} + V_{SS}}{2} \leq V_{DD} \\ V_{DD} &< A(U_+ - U_-) + \frac{V_{DD} + V_{SS}}{2} \end{aligned}$$



$A = \infty$   
IDEAL

$$\begin{aligned} * U^+ &< U^- & U_{out} &= V_{SS} \\ U^+ &> U^- & U_{out} &= V_{DD} \end{aligned}$$

## Charge Sharing

\* Floating node where charge can't flow in/out

\* Voltage drops from plate w/ positive charge to plate holding negative charge

① Label voltages across all capacitors

② Draw circuit in each phase

③ ID all floating nodes during phase 2

④ Examine each floating node individually

a) ID capacitor plates connected to that node phase 2

b) calculate charge on those plates phase 1

\* use node voltages according to labelled polarities

\* DON'T use parallel/series eq cap

\* Floating node  $\emptyset_2$  not always floating  $\emptyset_1$

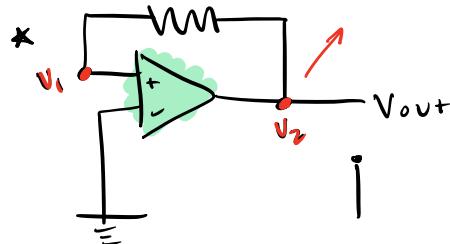
⑤ Find total charge on floating node in phase 2

⑥ Charge in steady state phase 1 = charge in phase 2

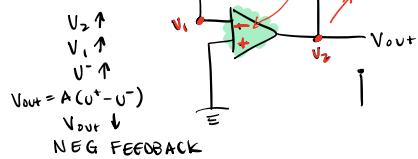
# NEG FEEDBACK

## GOLDEN RULES

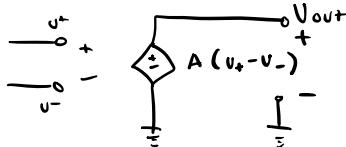
- ①  $I_+ = I_- = 0$  ← even w/o neg feedback
- ②  $V^+ = V^-$



$v_2 \uparrow$  then  $v_1 \uparrow$  so  $v^+ \uparrow$  and since  $V_{out} = A(v^+ - v^-)$   
so since  $v^+ \uparrow$  then  $V_{out} \uparrow$   
POSITIVE FEEDBACK!  
↳ FLIP polarities



SOME FUNCTION OF OUTPUT fed back to input  
TO KEEP OUTPUT AT SOME FINITE VALUE



$$\text{① GR2} \rightarrow \begin{cases} v^+ = v^- \\ v^- = v^+ = 0 \end{cases}$$

$$\text{② GR1} \rightarrow I_1 + I_2 = J^0$$

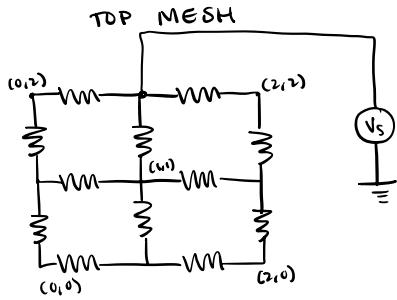
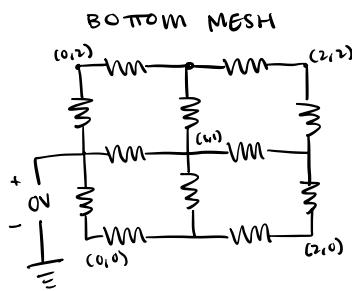
& NOW ADD KCL! & APPLY VOLTAGE DEFNS

$$\frac{v_2 - v_1}{R_1} + \frac{v_3 - v_2}{R_2} = 0$$

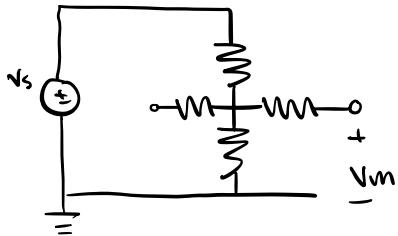
$$-\frac{v_1}{R_1} + \frac{v_2}{R_2} = 0$$

$$-\frac{v_{in}}{R_1} + \frac{V_{out}}{R_2} = 0$$

## 2D TOUCHSCREEN



## GENERAL CIRCUIT



# Design Procedure

- ① Restate goals of circuit
- ② Strategy: what you can measure, how/what you need to change (\*use block diagrams)
- ③ Implement: use blocks & think abt how they can be modified/extended
- ④ Verify & check block-to-block connections (check contradictions)

\* think about which elements depend on which aspects

Elements: op-amps, resistors, capacitors, comparators, switches

\* Use buffer to connect parts

\* Gain:  $A_v = \frac{\text{output voltage}}{\text{input voltage}}$

## Inner Product • Norms

$$\begin{aligned} \langle \vec{x}, \vec{y} \rangle &= \vec{x} \cdot \vec{y} = \left[ \begin{array}{c} \vec{x}^T \\ \vec{y} \end{array} \right] = [x_1, x_2, \dots, x_n] \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix} = x_1 y_1 + x_2 y_2 + \dots + x_n y_n \\ &= \|\vec{x}\| \|\vec{y}\| \cos \theta \end{aligned}$$

|orthogonal: when inner product = 0

Euclidean norm:  $\|\vec{x}\| = \sqrt{x_1^2 + x_2^2 + \dots + x_n^2} = \sqrt{\langle \vec{x}, \vec{x} \rangle}$

$\|\vec{x}\| = 0$  iff  $\vec{x} = \vec{0}$

$\|\alpha \vec{x}\| = |\alpha| \|\vec{x}\| \quad \|\vec{x} + \vec{y}\| \leq \|\vec{x}\| + \|\vec{y}\|$

NORM

(magnitude of vector)

① symmetry:  $\langle \vec{v}, \vec{v} \rangle = \langle \vec{v}, \vec{v} \rangle$

② linearity:  $\langle \alpha \vec{v}, \vec{v} \rangle = \alpha \langle \vec{v}, \vec{v} \rangle$

$\langle \vec{v} + \vec{w}, \vec{v} \rangle = \langle \vec{v}, \vec{v} \rangle + \langle \vec{w}, \vec{v} \rangle$

③ positive-definiteness:  $\langle \vec{v}, \vec{v} \rangle \geq 0$

$$\hat{x} = \frac{\vec{x}}{\|\vec{x}\|} \quad \hat{y} = \frac{\vec{y}}{\|\vec{y}\|}$$

$$\langle \vec{x}, \vec{x} \rangle = \|\vec{x}\|^2 = \vec{x}^T \vec{x}$$

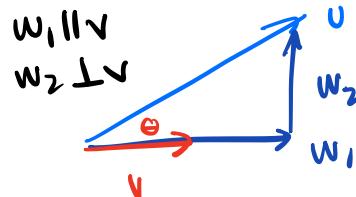
$$|\langle \vec{x}, \vec{y} \rangle| \leq \|\vec{x}\| \|\vec{y}\|$$

projection of  $\vec{x}$  onto  $\vec{y}$ :  $\text{proj}_{\vec{y}} \vec{x} = \frac{\langle \vec{y}, \vec{x} \rangle}{\|\vec{y}\|^2} \vec{y}$

$$\vec{e} = \vec{x} - \text{proj}_{\vec{y}} \vec{x} \rightarrow \langle \vec{e}, \vec{y} \rangle = \langle \vec{x} - \text{proj}_{\vec{y}} \vec{x}, \vec{y} \rangle = \langle \vec{x}, \vec{y} \rangle - \langle \vec{y}, \vec{y} \rangle = 0 \rightarrow \vec{e} \perp \vec{y}$$
 are orthogonal

$\vec{x} \in \mathbb{R}^n$  such that projection of  $\vec{b}$  onto col space A is  $A\vec{x}$  where x from least square

- \* vector  $\text{proj}_{\vec{y}} \vec{x}$  is vector in  $\text{span}\{\vec{y}\}$  that is closest to  $\vec{x}$
- \* BLG inner product = MORE similar



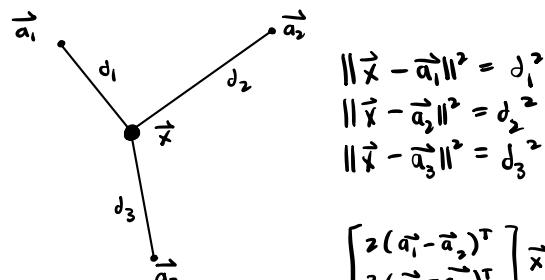
component of  $u$  that travels along  $v$  (proj of  $u$  onto  $v$ )

$$w_1 = \text{proj}_v u$$

$$w_2 = u - w_1 = u - \text{proj}_v u$$

orthogonal to  $v$

## Trilateration



$$\begin{aligned} (\vec{x} - \vec{a}_1)^T (\vec{x} - \vec{a}_1) &= \vec{x}^T \vec{x} - 2\vec{a}_1^T \vec{x} + \|\vec{a}_1\|^2 = d_1^2 \\ \vec{x}^T \vec{x} - 2\vec{a}_2^T \vec{x} + \|\vec{a}_2\|^2 &= d_2^2 \\ \vec{x}^T \vec{x} - 2\vec{a}_3^T \vec{x} + \|\vec{a}_3\|^2 &= d_3^2 \end{aligned}$$

$$\begin{bmatrix} 2(\vec{a}_1 - \vec{a}_2)^T \\ 2(\vec{a}_1 - \vec{a}_3)^T \end{bmatrix} \vec{x} = \begin{bmatrix} \|\vec{a}_1\|^2 - \|\vec{a}_2\|^2 - d_1^2 + d_2^2 \\ \|\vec{a}_1\|^2 - \|\vec{a}_3\|^2 - d_1^2 + d_3^2 \end{bmatrix}$$

\* subtract eqs to get rid of the quadratic term

## Cross-correlation

$$\text{corr}_{\vec{x}}(\vec{y}) [k] = \sum_{i=-\infty}^{\infty} x[i] y[i-k]$$

$$\text{circcorr}(\vec{x}, \vec{y}) [k] = \sum_{i=0}^{N-1} x[i] y[(i-k)_N]$$

$$\text{corr}_N(\vec{x}, \vec{y}) [k] = \sum_{i=0}^{N-1} x[i] y[i-k]$$

linear

circular

periodic

\* measure of similarity based on inner product

$$d = \sqrt{\tau}$$

$$\tau = \arg \max (\text{circorr}(\vec{r}, \vec{s})) [k]$$

\* received signal = same / other signal shifts

\* use zero padding

\* length-1 shifts to right & left

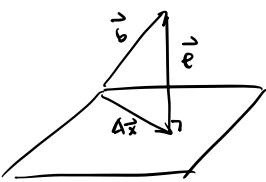
\* assume signal repeating w/ per N

auto-correlation: correlation between signal & itself:  $\text{corr}_{\vec{x}} \vec{x}$

$$\text{corr}_{\vec{y}}(\vec{x}) \neq \text{corr}_{\vec{x}}(\vec{y})$$

# Least Squares

$\vec{A}\vec{x} = \vec{b}$  where more equations than unknowns  
 $\|\vec{e}\| = \|\vec{b} - \vec{A}\vec{x}\|$



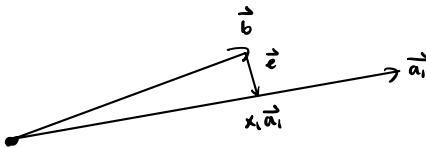
$$\langle \vec{e}, \vec{a}_i \rangle = 0$$

$$x_i = \frac{\langle \vec{b}, \vec{a}_i \rangle}{\langle \vec{a}_i, \vec{a}_i \rangle}$$

$$\|\vec{e}\| = \|\vec{b} - \vec{A}\vec{x}\|$$

↑ actual      ↑ predicted

$$\langle \vec{e}, \vec{a}_i \rangle = 0 \iff \vec{a}_i^T \vec{e} = 0 \rightarrow \vec{A}^T \vec{e} = \vec{0}$$



- \* NEED linearly indep columns
- \* rows  $\geq$  cols
- \* careful about actually applying model

$$\vec{x} = (\vec{A}^T \vec{A})^{-1} \vec{A}^T \vec{b}$$

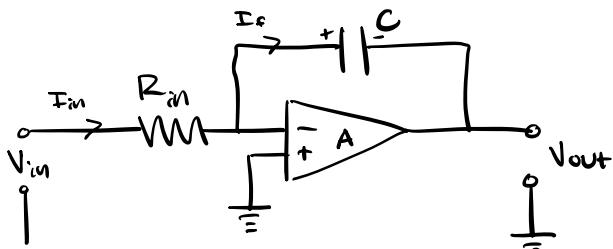
$\vec{e} = \vec{b} - \vec{A}\vec{x}$  is orthogonal to cols of  $\vec{A}$

$$\begin{aligned} (\vec{A}^T)(\vec{b} - \vec{A}\vec{x}) &= \vec{A}^T(\vec{b} - \vec{A}(\vec{A}^T \vec{A})^{-1} \vec{A}^T \vec{b}) \\ &= \vec{A}^T \vec{b} - \vec{A}^T \vec{A}(\vec{A}^T \vec{A})^{-1} \vec{A}^T \vec{b} \\ &= \vec{A}^T \vec{b} - \vec{I} \vec{A}^T \vec{b} \\ &= \vec{A}^T \vec{b} - \vec{A}^T \vec{b} = \vec{0} \end{aligned}$$

If vector is orthogonal to col( $\vec{A}$ ) it's in Null( $\vec{A}^T$ )

$$\langle \vec{e}, \vec{a}_i \rangle = 0 \iff \vec{a}_i^T \vec{e} = 0$$

↑ cols of  $\vec{A}$



\* Golden rules:

$$I_{in}(t) = \frac{V_{in}(t) - 0}{R_{in}} = \frac{V_{in}(t)}{R_{in}}$$

$$I_c(t) = I_{in}(t) = \frac{V_{in}(t)}{R_{in}}$$

$$V_{out}(t) = -\frac{1}{R_{in} C_{pix}} \int_0^t V_{in}(t') dt'$$

\* Current voltage for capacitors

$$I_c(t) = C_{pix} \frac{dV_c(t)}{dt}$$

$$V_c(t) = V_c(t_0) + \int_{t_0}^t \frac{I_c}{C_{pix}} dt$$

$$V_c(t) = V_c(t_0) + \int_{t_0}^t \frac{V_{in}(t')}{R_{in} C_{pix}} dt'$$