

Note 1 (Systems of Linear Equations)

- Linear Function:

$$f(\alpha x_1 + \beta y_1, \alpha x_2 + \beta y_2, \dots, \alpha x_n + \beta y_n) \\ \implies \alpha f(x_1, x_2, \dots, x_n) + \beta f(y_1, y_2, \dots, y_n)$$

- If $f: \mathbb{R}^n \rightarrow \mathbb{R}$ is linear, then

$$f(x_1, x_2, \dots, x_n) = c_1 x_1 + c_2 x_2 + \dots + c_n x_n$$

- Affine function:

$$g(x_1, x_2, \dots, x_n) = f(x_1, x_2, \dots, x_n) + c_0$$

for a linear function $f: \mathbb{R}^n \rightarrow \mathbb{R}$ and constant $c_0 \in \mathbb{R}$.

Note 2 (Vectors/Matrices)

- Vectors $\vec{x} \in \mathbb{R}^n$

- each x_i = component/element
- size = # of elements (n)
- vectors are equal if same size and elements are equal

- Standard unit vector: all elements 0 except one 1

- Vector multiplication:

- $\vec{y}^T \vec{x}$ = dot product; $\sum x_i y_i$

$$-\vec{x} \vec{y}^T = \text{matrix: } \begin{bmatrix} x_1 y_1 & x_1 y_2 & \dots \\ x_2 y_1 & \ddots & \vdots \\ \vdots & \dots & x_n y_n \end{bmatrix}$$

- Matrix multiplication:

- **AB**: for each row of **A**, multiply and sum for each col of **B**
- Associative, not commutative:
(**AB**)**C** = **A**(**BC**); **AB** \neq **BA**

- Identity matrix: **I**; 1's along diagonal, 0's everywhere else

Note 3 (Linear Independence/Span)

- Set of vectors $\{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_n\}$ is linearly dependent (LD) if $\alpha_1 \vec{v}_1 + \alpha_2 \vec{v}_2 + \dots + \alpha_n \vec{v}_n = \vec{0}$, where some $\alpha_i \neq 0$
- Or, LD if \vec{v}_i can be written as $\sum \alpha_j \vec{v}_j$, where some $\alpha_j \neq 0$
- Linearly independent (LI) if $\sum \alpha_i \vec{v}_i = \vec{0}$ only if all $\alpha_i = 0$
- Span: set of all linear combinations of the vectors

Note 5 (Water Pumps)

- Transition matrix (flow from cols to rows):

$$\begin{array}{c} \begin{array}{ccc} & \begin{array}{c} A \quad B \quad C \end{array} \\ \begin{array}{c} A \\ B \\ C \end{array} & \begin{bmatrix} A \rightarrow A & B \rightarrow A & C \rightarrow A \\ A \rightarrow B & B \rightarrow B & C \rightarrow B \\ A \rightarrow C & B \rightarrow C & C \rightarrow C \end{bmatrix} \end{array}$$

- Columns sum to 1 \implies conservative system (everything goes somewhere)

Note 6 (Matrix Inversion)

- A** is invertible if there exists a **B** s.t. **AB** = **BA** = **I**
- Finding inverse: Gaussian elimination, like solving **AX** = **I**; i.e. reduce $[\mathbf{A} \mid \mathbf{I}] \rightarrow [\mathbf{I} \mid \mathbf{A}^{-1}]$
- If invertible, then: (baby version of IMT)
 - rows and cols LI
 - $\mathbf{A}\vec{x} = \vec{b}$ has a unique solution for all \vec{b}
 - has a trivial nullspace
 - $\det \mathbf{A} \neq 0$

Note 7 (Vector Spaces)

- V is a vector space if:
 - $\vec{0} \in V$
 - $\vec{x}, \vec{y} \in V$ implies $\alpha \vec{x} + \beta \vec{y} \in V$ for all $\alpha, \beta \in \mathbb{R}$; i.e.
 - * closed under vector addition (if $\vec{x}, \vec{y} \in V$ then $\vec{x} + \vec{y} \in V$)
 - * closed under scalar multiplication (if $\vec{x} \in V$ then $\alpha \vec{x} \in V$ for all $\alpha \in \mathbb{R}$)
 - (among other things, but these are the most important)
- Basis of a vector space:
 - LI vectors, can express any $\vec{v} \in V$ as a linear combination of basis vectors, and is a minimum set of vectors that does so (implied by being LI)
- Dimension of a vector space = # of basis vectors
- All equivalent bases of a vector space must have the same dimension

Note 8 (Matrix subspaces)

- U is a subspace if it satisfies the 3 points in Note 7 (a subspace is a subset of a vector space)
- Column space: $\text{range}(\mathbf{A}) = \text{span}(\mathbf{A}) = \mathbf{C}(\mathbf{A}) = \text{span of cols of } \mathbf{A}$
- $\text{rank}(\mathbf{A}) = \dim(\text{span}(\mathbf{A}))$
- Nullspace: $\text{Null}(\mathbf{A}) = \mathbf{N}(\mathbf{A}) = \text{set of all } \vec{x} \text{ s.t. } \mathbf{A}\vec{x} = \vec{0}$
- nullity(**A**) = $\dim(\mathbf{N}(\mathbf{A}))$
- rank-nullity theorem: $\text{rank}(\mathbf{A}) + \text{nullity}(\mathbf{A}) = \# \text{ of columns of } \mathbf{A}$

Note 9 (Eigenvalues/Eigenvectors)

- If $\mathbf{A}\vec{x} = \lambda \vec{x}$, then \vec{x} is an eigenvector, λ is an eigenvalue of **A**
- Calculating eigenvalues: solve $\det(\mathbf{A} - \lambda \mathbf{I}) = 0$ for λ
- Calculating eigenvectors: solve $(\mathbf{A} - \lambda \mathbf{I})\vec{v} = \vec{0}$ for \vec{v}
- Repeated eigenvalues: multiple eigenvectors, same eigenvalue; forms an eigenspace
- If $(\lambda_1, \vec{v}_1), (\lambda_2, \vec{v}_2)$ are two distinct eigenpairs, then \vec{v}_1, \vec{v}_2 are LI.
- Characteristic polynomial: $\det(\mathbf{A} - \lambda \mathbf{I})$
- Steady states (water pumps): \vec{x} s.t. $\mathbf{A}\vec{x} = \vec{x}$ (i.e. eigenspace for $\lambda = 1$)
- $\lim_{n \rightarrow \infty} \mathbf{A}^n \vec{x} = \lim_{n \rightarrow \infty} \lambda^n \vec{x}$ if (λ, \vec{x}) is an eigenpair of **A**

Note 10 (Change of Basis/Diagonalization)

- If $\mathbf{T}\vec{u} = \vec{v}$, and \vec{u}_A and \vec{v}_B are vectors in the **A** and **B** bases respectively (i.e. columns of **A** and **B** are basis vectors in the new coordinate system), then arrows represent consecutive left-multiplication:

$$\begin{array}{ccc} \vec{u} & \xrightarrow{\mathbf{T}} & \vec{v} \\ \mathbf{A}^{-1} \left(\downarrow \right) \mathbf{A} & & \mathbf{B} \left(\downarrow \right) \mathbf{B}^{-1} \\ \vec{u}_A & \xrightarrow{\mathbf{B}^{-1} \mathbf{T} \mathbf{A}} & \vec{v}_B \end{array}$$

- Diagonalization: $\mathbf{A} = \mathbf{P} \mathbf{D} \mathbf{P}^{-1}$; **P** = matrix of eigenvectors, **D** = diagonal matrix of eigenvalues such that $\mathbf{A}\vec{v}_i = \lambda_i \vec{v}_i$

$$\mathbf{P} = \begin{bmatrix} | & & | \\ \vec{v}_1 & \dots & \vec{v}_n \\ | & & | \end{bmatrix} \quad \mathbf{D} = \begin{bmatrix} \lambda_1 & 0 & 0 \\ 0 & \ddots & 0 \\ 0 & 0 & \lambda_n \end{bmatrix}$$

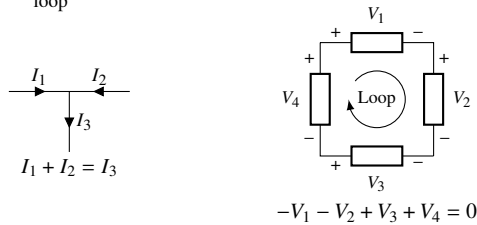
- Only diagonalizable if eigenvalues are linearly independent (i.e. if all eigenvalues are distinct)

Other

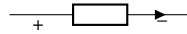
- $(\mathbf{AB})^T = \mathbf{B}^T \mathbf{A}^T$
- rotation matrix by θ counterclockwise: $\begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$
- $\det \begin{bmatrix} a & b \\ c & d \end{bmatrix} = ad - bc$
- $\begin{bmatrix} a & b \\ c & d \end{bmatrix}^{-1} = \frac{1}{ad - bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$
- $\det(\mathbf{AB}) = \det(\mathbf{A}) \det(\mathbf{B})$
- $\det(\mathbf{A}^{-1}) = \det(\mathbf{A})^{-1}$
- $\det(\mathbf{A})$ = product along diagonal if **A** is triangular
 - eigenvalues of a triangular matrix are the values along its diagonal

Note 11 (Circuits)

- Ohm's law: $V = IR$
- KCL: $I_{in} = I_{out}$
- KVL: $\sum_{loop} V_k = 0$; $- \rightarrow + = \text{add}$, $+ \rightarrow - = \text{subtract}$



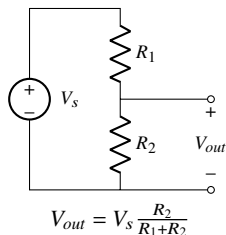
- NVA:
 - Label everything
 - Passive sign convention: current goes into +, out of -



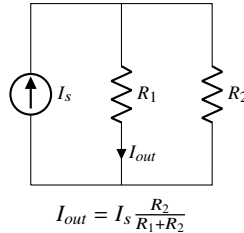
- Write KCL at each unknown node
- Substitute Ohm's law for each current
- Solve for desired values

Note 12 (Resistive Touchscreen)

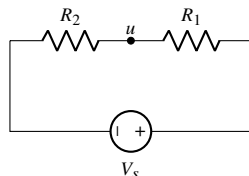
- Voltage divider:



- Current divider:



- $R = \rho \frac{L}{A} = \rho \frac{\text{length}}{\text{area}}$, where ρ = resistivity
- Resistive touchscreen: touch splits resistor



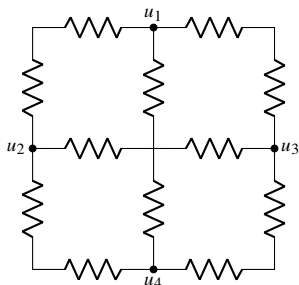
$$R_2 = \rho \frac{L_{touch}}{A} \quad R_1 = \rho \frac{L - L_{touch}}{A} \quad u = V_s \frac{L_{touch}}{L}$$

Note 13 (Power)

- Power: $P = VI = \frac{V^2}{R} = I^2 R$
- Voltmeter: connected *in parallel* to element to measure voltage drop
- Ammeter: embedded in the circuit *in series* to measure current

Note 14 (2D Resistive Touchscreen)

- 2D resistive touchscreen



- Powering $u_1 \rightarrow u_4$: measure y position

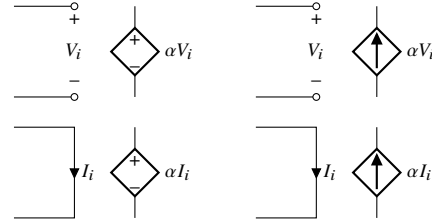
$$V_{out} = V_s \frac{L_{touch, vertical}}{L}$$

- Powering $u_2 \rightarrow u_3$: measure x position

$$V_{out} = V_s \frac{L_{touch, horizontal}}{L}$$

Note 15 (Superposition, Equivalences)

- Dependent sources:



- Superposition:

- for each independent source:
 - replace voltage source with wire, current source with open circuit
 - leave everything alone, find value (keep the same signs!)
- sum up everything

- Resistor equivalences:

- Parallel: $R_{eq} = \frac{R_1 R_2}{R_1 + R_2}$
- Series: $R_{eq} = R_1 + R_2$

- Voltage drop is equal through parallel branches (adjacent to same nodes)

- Current is equal through elements in series (by KCL)

Note 16 (Capacitors)

- Capacitors:

- charge (on positive plate) = $Q = CV$; C = capacitance
- $I = C \frac{dV}{dt}$
- if constant current, then $I = C \frac{\Delta V}{\Delta t}$ and $It = C(V(t) - V(0))$

- Capacitor equivalences:

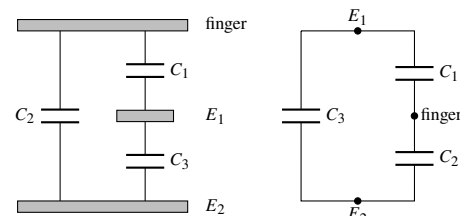
- Parallel: $C_{eq} = C_1 + C_2$
- Series: $C_{eq} = \frac{C_1 C_2}{C_1 + C_2}$

- $C = \epsilon \frac{A}{d} = \epsilon \frac{\text{area}}{\text{distance}}$, where ϵ = permittivity

- Energy: $E = \frac{1}{2} CV^2$

Note 17 (Capacitive Touchscreen)

- Capacitive touch screen:



- Touch adds parallel capacitors (C_1, C_2) \Rightarrow increased capacitance

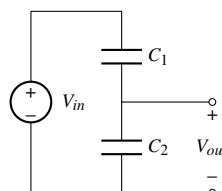
Note 17B (Charge Sharing)

- Charge sharing steps:

- Draw/label phases, keep polarity/signs for elements consistent through phases
- For all floating nodes in phase 2, use charge conservation; find total charge on adjacent plates (keep + and - plates in mind!)
- Equate with the total charge on the *same* plates in phase 1

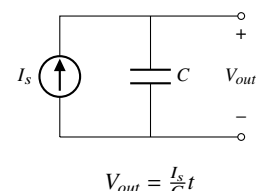
Other

- Capacitive divider:



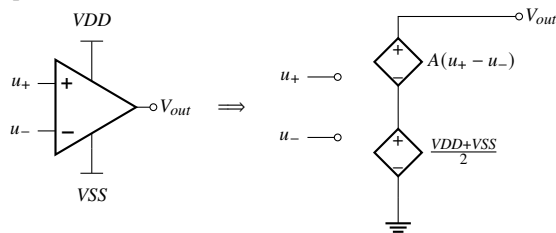
$$V_{out} = V_{in} \frac{C_1}{C_1 + C_2}$$

- Charging a capacitor:



Note 18/19 (Op Amps)

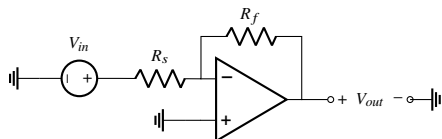
- Op amp:



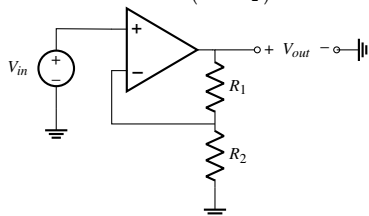
- Ideal op amp:

- $A \rightarrow \infty$
- No current through u_+ , u_-
- $u_+ - u_- = 0$

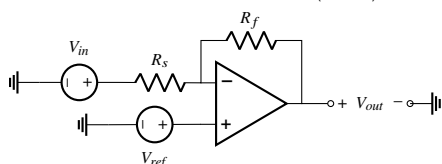
- Inverting Amplifier: $V_{out} = V_{in} \left(-\frac{R_f}{R_s} \right)$



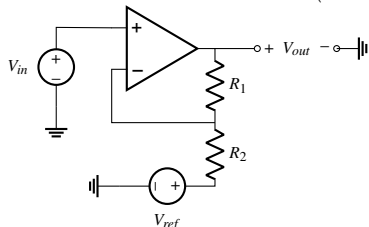
- Noninverting Amplifier: $V_{out} = V_{in} \left(1 + \frac{R_1}{R_2} \right)$



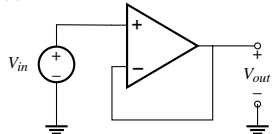
- Inverting Amplifier w/ reference: $V_{out} = V_{in} \left(-\frac{R_f}{R_s} \right) + V_{ref} \left(1 + \frac{R_f}{R_s} \right)$



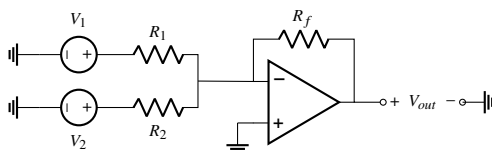
- Noninverting Amplifier w/ reference: $V_{out} = V_{in} \left(1 + \frac{R_1}{R_2} \right) - V_{ref} \left(\frac{R_1}{R_2} \right)$



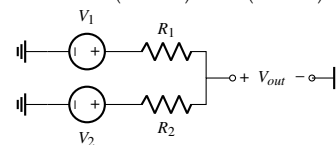
- Unity Gain Buffer: $V_{in} = V_{out}$



- Inverting Summing Amplifier: $V_{out} = -R_f \left(\frac{V_{in1}}{R_1} + \frac{V_{in2}}{R_2} \right)$



- Voltage Summer: $V_{out} = V_1 \left(\frac{R_2}{R_1+R_2} \right) + V_2 \left(\frac{R_1}{R_1+R_2} \right)$

**Note 21 (Inner Products)**

- (Euclidean) Inner product: $\langle \vec{x}, \vec{y} \rangle = \vec{x} \cdot \vec{y} = \vec{x}^T \vec{y} = \sum x_i y_i$
- \vec{x}, \vec{y} are orthogonal if $\langle \vec{x}, \vec{y} \rangle = 0$
- $\langle a\vec{x}, \vec{y} \rangle = a \langle \vec{x}, \vec{y} \rangle$ and $\langle \vec{x} + \vec{y}, \vec{z} \rangle = \langle \vec{x}, \vec{z} \rangle + \langle \vec{y}, \vec{z} \rangle$
- Norm: $\|\vec{x}\| = \sqrt{\langle \vec{x}, \vec{x} \rangle}$ = length/magnitude of vector
- Alternate definition: $\langle \vec{x}, \vec{y} \rangle = \|\vec{x}\| \|\vec{y}\| \cos \theta$
- Cauchy-Schwarz inequality: $|\langle \vec{x}, \vec{y} \rangle| \leq \|\vec{x}\| \|\vec{y}\|$

Note 22 (Correlation)

- Cross-correlation: $\text{corr}_{\vec{x}}(\vec{y})[k] = \sum_{i=-\infty}^{\infty} x[i]y[i-k]$
- $\vec{x}[i], \vec{y}[i] = 0$ outside of defined range
- $\text{corr}_{\vec{x}}(\vec{y})[k] = \text{corr}_{\vec{y}}(\vec{x})[-k]$; they're mirrored
- Autocorrelation: $\text{corr}_{\vec{x}}(\vec{x})$
- Circular correlation:

$$\text{circcorr}(\vec{x}, \vec{y}) = \begin{bmatrix} \text{---} & \text{rows are all} & \text{---} \\ \text{---} & \text{circular shifts} & \text{---} \\ \text{---} & \text{of } \vec{y} & \text{---} \end{bmatrix} \vec{x}$$

Note 23 (Projection/Least Squares)

- Projection of \vec{b} onto \vec{a} : $\text{proj}_{\vec{a}}(\vec{b}) = \frac{\langle \vec{b}, \vec{a} \rangle}{\langle \vec{a}, \vec{a} \rangle} \vec{a}$
- Scalar projection of \vec{b} onto \vec{a} : $\frac{\langle \vec{b}, \vec{a} \rangle}{\langle \vec{a}, \vec{a} \rangle}$
- Projection onto subspace: if columns of \mathbf{A} are orthogonal, then $\text{proj}_{\mathbf{A}}(\vec{b}) = \sum \text{proj}_{\vec{a}_i}(\vec{b})$ where \vec{a}_i are columns of \mathbf{A} ; if not, use least squares
- Least squares: to minimize the error $e = \|\mathbf{A}\vec{x} - \vec{b}\|$, we have $\hat{\vec{x}} = (\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T \vec{b}$
- Setting up least squares:
 - \mathbf{A} = matrix of known values/coefficients
 - \vec{x} = vector of variables
 - \vec{b} = vector of constants
- $\mathbf{A}^T \mathbf{A}$ is invertible if \mathbf{A} has LI columns (i.e. can only apply least squares if \mathbf{A} has LI columns)

Other

- Trilateration:
 - n variables $\implies n$ equations if linear, $n+1$ equations if nonlinear (subtract from one equation to linearize)
 - in space: n dimensions $\implies n+1$ equations for circles/spheres; one is sacrificed to linearize
 - if delays are unknown, need $n+2$ equations; sacrifice one for reference, sacrifice another to linearize
- Units (good to double check calculations)
 - Current: $A = C/s = \text{charge/time}$
 - Voltage: $V = J/C = \text{energy/charge}$
 - Resistance: $\Omega = V/A$
 - Power: $W = J/s = \text{energy/time}$
 - Capacitance: $F = C/V = \text{charge/volt}$