

The background of the slide is a detailed microchip layout. It features a grid of small squares representing the chip's surface. Overlaid on this grid are various circuit components and interconnects. Several components are highlighted with dashed yellow boxes and labeled in yellow text: 'RX' (Receiver), 'LO Buffer' (Local Oscillator Buffer), 'Hybrid', 'Wilkinson' (referring to a Wilkinson power divider), another 'LO Buffer', and 'TX' (Transmitter). The layout is complex, showing a dense network of lines and blocks representing the physical design of the chip.

# EECS 16B

## Designing Information Devices and Systems II

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# Module 6: Change of Basis

EECS 16B

# Where do you live?

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- In a 2D plane, we only need to specify three points:
  - Origin (Downtown Library)
  - Location 1 (Cream on Telegraph)
  - Location 2 (Cory Hall)
- Equivalently, we need to specify two vectors
- As long as the vectors are not co-linear, we can specify any other location
- In other words, the vectors must be linearly independent
  - $\vec{v}_1 \neq a\vec{v}_2$

# Generalize to 3D and Beyond

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- Any  $N$  vectors that are linearly independent can form a basis in an  $N$  dimensional space.

# Standard Basis

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- $\hat{e}_1, \hat{e}_2, \dots, \hat{e}_N$  are just one representation of mutually orthogonal vectors.
- Note that even the standard basis is not unique as we can easily rotate the standard basis to find a new basis



# Sather Gate

- Here we take  $\hat{e}_1$  and  $\hat{e}_2$  to be North and East
- Which directions are better?
- Both get you to the same location, so they are equivalent.



# How to change basis?

- Say you know a location using basis set  $\vec{v}_1, \vec{v}_2, \dots$  but you're talking to someone who has no idea where Cory Hall or another other place in Berkeley resides
- What to do? Tell them to pull out a compass and walk along  $e_1, e_2$  instead
- Since the destination is the same, we have:

$$\vec{z} = x_1 \hat{e}_1 + x_2 \hat{e}_2 = y_1 \vec{v}_1 + y_2 \vec{v}_2$$

$\vec{v}_1 \neq k \vec{v}_2$        $\vec{v}_1 \& \vec{v}_2$  linearly independent

$$\lambda_1 \hat{e}_1 + \lambda_2 \hat{e}_2 + \dots + \lambda_n \hat{e}_n = I \vec{x}$$

$$I \vec{x} = \begin{bmatrix} 1 & 0 & & 0 \\ 0 & 1 & & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & & 1 \end{bmatrix} = I \vec{x}$$

$$I \vec{x} = y_1 \vec{v}_1 + y_2 \vec{v}_2 + \dots + y_n \vec{v}_n$$

$$I \vec{x} = V \vec{y} \quad \boxed{\vec{y} = V^{-1} \vec{x}}$$

Transformation Matrix



$$\vec{z} = A \vec{x}_A = B \vec{x}_B$$

$$A = [\vec{a}_1 \ \vec{a}_2 \ \dots \ \vec{a}_n]$$

$$B = [\vec{b}_1 \ \vec{b}_2 \ \dots \ \vec{b}_n]$$

$$\begin{aligned} \vec{x}_B &= \underbrace{B^{-1} A}_{T_{BA}} \vec{x}_A \\ &= T_{BA} \vec{x}_A \end{aligned}$$

# Matrix Product: Standard Basis

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- Recall that a matrix-vector multiplication is just a linear combination of the columns of a matrix.
- If we form two matrices, we have:
- But  $E = I$ , so we have:
- Likewise, since  $V$  has independent columns

# Non-Standard Basis

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- In the previous calculation one basis was the standard basis
- What if we have two non-standard basis vectors  $A$  and  $B$ 
  - Why can we always invert  $A$ ?
    - The columns of  $A$  are linearly independent !

# Standard Basis to New Basis

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- We found that:
- If  $B$  is the standard basis, then  $B = I$  and we have

# Linear Transformations

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- Let's see how a linear transformation in one basis can be written in another basis
- Say  $L_{\mathcal{V}}$  is a linear transformation (mapping) from  $\vec{x}$  to  $\vec{y}$  both represented in basis  $A$
- In other words, for all  $x$  we have
- To represent  $L_{\mathcal{V}}$  as a matrix, we note that  $\vec{x}$  can be written as



$$\mathcal{L}\{\vec{v}\} \rightarrow \vec{z}$$

$$\vec{v} = x_1 \vec{v}_1 + x_2 \vec{v}_2 + x_3 \vec{v}_3 + \dots$$

$$\mathcal{L}\{\vec{v}\} = x_1 \underbrace{\mathcal{L}\{\vec{v}_1\}}_{\vec{a}_1} + x_2 \underbrace{\mathcal{L}\{\vec{v}_2\}}_{\vec{a}_2} + \dots$$

$$= x_1 \vec{a}_1 + x_2 \vec{a}_2 + x_3 \vec{a}_3 + \dots$$

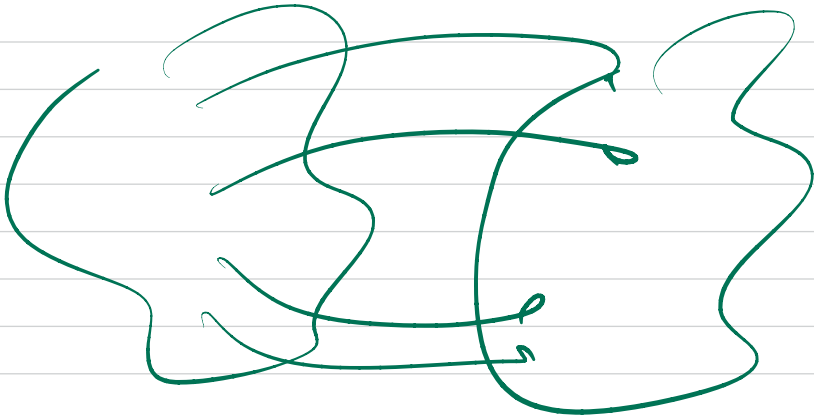
$$= \underline{A} \vec{x}$$

$$\mathcal{L}_{\vec{r}} \{ \} = V$$

$$\mathcal{L}_{\vec{y}} \{ \} = Y$$

$\mathcal{L}$  same mapping

$$\vec{z} = V \vec{v}$$



$$\vec{v}_1 = \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix}$$

$$\begin{pmatrix} 0 \\ \vdots \\ 0 \end{pmatrix} \neq \alpha \vec{v}_2$$

# Transformation Matrix

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- In other words, the matrix columns are the result of applying the transformation to  $\vec{a}_1, \vec{a}_2 \dots$
- We form a matrix of columns of  $L_A = L_v(\vec{a}_k)$

# Transformation in Another Basis

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- Now we can always use another basis  $B$ , and define  $L_B$  in the same way
- Since both transformations applied to the same input must map to the same point, we can say

$$\vec{y} = T \vec{x}$$

$T$  in standard basis

$$\vec{y}_B = T_B \vec{x}_B$$

$T$  &  $T_B$  should be related

# Review

$$\vec{b}_1, \vec{b}_2, \dots, \vec{b}_n$$

Standard basis  $\vec{x} = x_1 \hat{e}_1 + x_2 \hat{e}_2 + \dots$

$$B = [\vec{b}_1; \vec{b}_2; \dots; \vec{b}_n]$$

$$\vec{x}_B = B^{-1} \vec{x}$$



# Manipulating...

- Moving vectors from basis  $\overset{I}{X}$  to  $B$ :

$$\vec{y} = T \vec{x}$$

$$B^T \vec{y}_B = B^T T \vec{x}$$

- Since the matrix is invertible:

$$\vec{y}_B = B^{-T} T \vec{x}$$

$$\vec{x}_B = B^T \vec{x}$$

$$\vec{x} = (B \vec{x}_B)$$

- The mappings are therefore related as follows:

$$\vec{y}_B = \underbrace{B^T T B}_{T_B} \vec{x}_B$$

$T$  in standard basis

$$T_B = B^T T B$$

$T$  in basis  $B$

$T$  &  $T_B$  are "Similar Matrices"

$$T_B = B^{-1} T B$$

$$B T_B B^{-1} = T \quad \leftarrow$$

$$B T_B B^{-1} = T$$

# Diagonalizing Basis

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- Is there a natural basis ? Earlier we argued that all basis vectors were equivalent but with respect to a particular linear mapping, what's the best basis ?
- If a particular basis turns a transformation into a diagonal matrix, we say that the basis is the simplest basis for a transformation  $T$
- Recall the definition of an eigenvector:
  - $T\vec{v} = \lambda\vec{v}$       *By definition*
  - where the eigenvector is  $\vec{v}$  and eigenvalue is  $\lambda$

$$B^{-1} T B = D \quad \text{diagonal}$$



what basis  $B$  turns our transformation  
 $T$  into a diagonal matrix?

# Eigenvector Basis

- If the set of eigenvectors forms a basis (linearly independent), we know that the transformation is invertible, and we can use the eigenvectors as a basis:  $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_n$

- Let's perform a transformation using the eigenfunction basis vectors:

$$\begin{aligned} T \vec{a} &= T (u_1 \vec{v}_1 + u_2 \vec{v}_2 + \dots) \\ &= u_1 T \vec{v}_1 + u_2 T \vec{v}_2 + \dots \\ &= u_1 \lambda_1 \vec{v}_1 + u_2 \lambda_2 \vec{v}_2 + \dots \end{aligned} \quad \vec{a} = \begin{pmatrix} u_1 \\ u_2 \\ \vdots \\ u_n \end{pmatrix}$$

- We see the result is very simple as each vector is mapped to itself times a constant



# Matrix Form of Eigenvector Basis

- Let's rewrite:

$$T \vec{u} = u_1 \lambda_1 \vec{v}_1 + u_2 \lambda_2 \vec{v}_2 + u_3 \lambda_3 \vec{v}_3 + \dots$$

- As:  $Q = [ \vec{v}_1 \ \vec{v}_2 \ \dots ]$

$Q^{-1}$  transformation matrix from standard basis to eigen-basis

$$T \vec{a} = Q \Lambda \vec{a}$$
$$= Q \begin{bmatrix} \lambda_1 & & 0 \\ & \dots & \\ 0 & & \lambda_n \end{bmatrix} \vec{a}$$

- Here we form the eigenvector matrix  $Q$ , diagonal matrix  $\Lambda$

$$\Lambda \vec{u} = \begin{pmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{pmatrix} \begin{pmatrix} u_1 \\ \vdots \\ u_n \end{pmatrix} = \begin{pmatrix} \lambda_1 u_1 \\ \lambda_2 u_2 \\ \vdots \\ \lambda_n u_n \end{pmatrix}$$

$$\Phi \Lambda \vec{u} = \Phi \begin{pmatrix} \lambda_1 u_1 \\ \vdots \\ \lambda_n u_n \end{pmatrix}$$

$$= \vec{r}_1 \lambda_1 u_1 + \vec{r}_2 \lambda_2 u_2 + \dots$$

# Eigenvector Basis (cont)

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- So far, we have found that

$$T\vec{u} = Q\Lambda(\vec{u})$$

- Now let's write  $\vec{u}$  in terms of the standard basis:

$$\vec{u} = Q^T \vec{x}$$

- The transformation:  $T\vec{u} = TQ^T \vec{x} = \underbrace{Q\Lambda Q^T}_{\Lambda} \vec{x}$

In the new basis  $T$  is  $\Lambda$

# Deconvolving a Linear Mapping

- How is this useful? Suppose we want to solve

$$T \vec{x} = \vec{b} \quad Q \sim \underbrace{Q^{-1} T Q}_{\Lambda} \vec{x} = \vec{b}$$

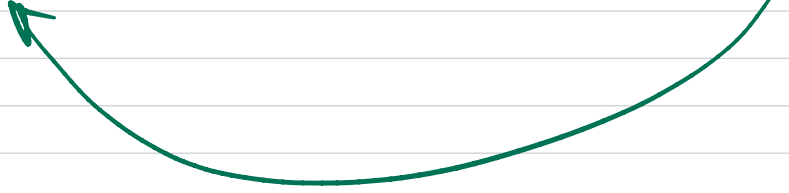
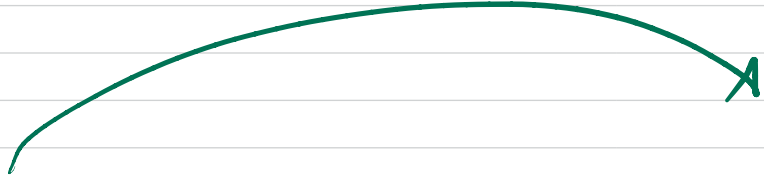
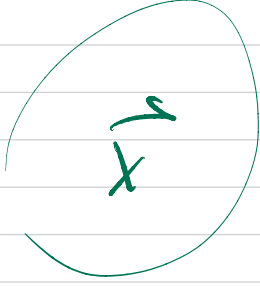
- Recall that  $\vec{u}$  is the vector  $\vec{u}$  in the basis of eigenvectors of  $T$

$$\vec{u} = Q^{-1} \vec{x} \quad Q \sim \Lambda \vec{u} = \vec{b}$$

- Like  $Q^{-1} \vec{b}$  is simply  $\vec{b}$  transformed into the eigenspace

$$\underbrace{Q^{-1} Q}_{I} \sim \Lambda \vec{u} = Q^{-1} \vec{b} \quad \vec{b} \text{ is } \vec{b} \text{ in eigen-basis}$$

$\Lambda \vec{u} = \vec{b}$



$\psi$

# The World's Easiest Matrix To Invert

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- If we know the eigenvectors of a linear transform, it can save us a ton of work.
- Instead of solving a difficult  $n \times n$  problem we solve just  $n$  equations:

$$\Lambda^{-1} = \begin{pmatrix} \lambda_1^{-1} & & 0 \\ & \ddots & \\ 0 & & \lambda_n^{-1} \end{pmatrix}$$

- $\Lambda$  is a diagonal matrix !
- Clearly the eigenvector basis is the most “natural” for solving problems involving a linear transform  $T$

# Use the best “map”

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- Summarizing what we have learned, we can say that if we change our perspective (basis) and view things in terms of the eigenspace, then a linear transformation is very simple and only involves multiplying by a diagonal matrix
- Can we apply this to circuit and other dynamic systems described by linear differential equations ... ?

# Power to the Matrix

- Consider the n'th power of a matrix:

$$A^n = A \cdots A = \underbrace{(Q \Lambda Q^{-1}) (Q \Lambda Q^{-1}) \cdots (Q \Lambda Q^{-1})}_I$$

$$= Q \Lambda I \Lambda Q^{-1} \cdots$$

$$= Q \Lambda^2 Q^{-1} \underbrace{(Q \Lambda Q^{-1}) \cdots (Q \Lambda Q^{-1})}_I = Q \Lambda^n Q^{-1}$$

- Note that raising a diagonal matrix is trivial!

$$= Q \Lambda^n Q^{-1}$$



# Preview: Frequency Domain Basis

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- We have found that complex exponentials pretty much solve any homogeneous constant coefficient differential equations
- These equations arise from time-invariant circuits and in general time-invariant systems
  - The component values don't change with time
- Another way to see this is to say that the complex exponential is an *eigenfunction* of a linear dynamic system !
- Does it form a basis ?
  - Yes, in fact it forms a continuous orthogonal basis

# Preview (cont)

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- The Fourier Transform and Laplace Transform (EE 120) take you from the time domain to the frequency domain, in other words they are a change of basis operation
- If we solve problems in the frequency domain, these complicated differential equations collapse into simple “diagonal” systems that we can solve ...
  - “Convolution” turns into multiplication