

# Fundamentals of Linear Algebra

Class 2-3. 1 Sep 2011

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## Administrivia

- Registration: Anyone on waitlist still?
- Homework 1: Will be handed out with class 3.
  - Linear algebra

## Overview

- Vectors and matrices
- Basic vector/matrix operations
- Vector products
- Matrix products
- Various matrix types
- Matrix inversion
- Matrix interpretation
- Eigenanalysis
- Singular value decomposition

## Book

- Fundamentals of Linear Algebra, Gilbert Strang
- Important to be very comfortable with linear algebra
  - Appears repeatedly in the form of Eigen analysis, SVD, Factor analysis
  - Appears through various properties of matrices that are used in machine learning, particularly when applied to images and sound
- Today's lecture: Definitions
  - Very small subset of all that's used
  - Important subset, intended to help you recollect

## Incentive to use linear algebra

- Pretty notation!

$$\mathbf{x}^T \cdot \mathbf{A} \cdot \mathbf{y} \longleftrightarrow \sum_j y_j \sum_i x_i A_{ij}$$

- Easier intuition

- Really convenient geometric interpretations
- Operations easy to describe verbally

- Easy code translation!

```

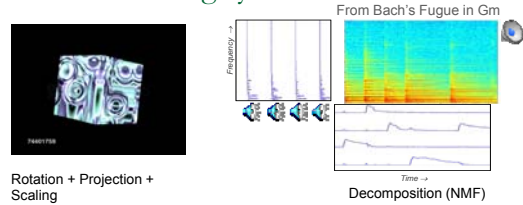
for i=1:n
    for j=1:m
        c(i)=c(i)+y(j)*x(i)*a(i,j)
    end
end
    
```

 $\longleftrightarrow$ 

```

C=x*A*y
    
```

## And other things you can do



- Manipulate Images
- Manipulate Sounds

## Scalars, vectors, matrices, ...

- A *scalar*  $a$  is a number
  - $a = 2$ ,  $a = 3.14$ ,  $a = -1000$ , etc.
- A *vector*  $\mathbf{a}$  is a linear arrangement of a collection of scalars

$$\mathbf{a} = [1 \quad 2 \quad 3] \quad \mathbf{a} = \begin{bmatrix} 3.14 \\ -32 \end{bmatrix}$$

- A *matrix*  $\mathbf{A}$  is a rectangular arrangement of a collection of vectors

$$\mathbf{A} = \begin{bmatrix} 3.12 & -10 \\ 10.0 & 2 \end{bmatrix}$$

- MATLAB syntax:  $\mathbf{a} = [1 \ 2 \ 3]$ ,  $\mathbf{A} = [1 \ 2; 3 \ 4]$

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## Vector/Matrix types and shapes

- Vectors are either column or row vectors

$$\mathbf{c} = \begin{bmatrix} a \\ b \\ c \end{bmatrix}, \mathbf{r} = [a \ b \ c] \quad \mathbf{s} = [\text{waveform}]$$

- A sound can be a vector, a series of daily temperatures can be a vector, etc ...
- Matrices can be square or rectangular

$$\mathbf{S} = \begin{bmatrix} a & b \\ c & d \end{bmatrix}, \mathbf{R} = \begin{bmatrix} a & b & c \\ d & e & f \end{bmatrix}, \mathbf{M} = \text{[image]}$$

- Images can be a matrix, collections of sounds can be a matrix, etc ...

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## Dimensions of a matrix

- The matrix size is specified by the number of rows and columns

$$\mathbf{c} = \begin{bmatrix} a \\ b \\ c \end{bmatrix}, \mathbf{r} = [a \ b \ c]$$

- $\mathbf{c} = 3 \times 1$  matrix: 3 rows and 1 column
- $\mathbf{r} = 1 \times 3$  matrix: 1 row and 3 columns

$$\mathbf{S} = \begin{bmatrix} a & b \\ c & d \end{bmatrix}, \mathbf{R} = \begin{bmatrix} a & b & c \\ d & e & f \end{bmatrix}$$

- $\mathbf{S} = 2 \times 2$  matrix
- $\mathbf{R} = 2 \times 3$  matrix
- Pacman =  $321 \times 399$  matrix

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## Representing an image as a matrix



- 3 pacmen

- A  $321 \times 399$  matrix
  - Row and Column = position
- A  $3 \times 128079$  matrix
  - Triples of  $x, y$  and value
- A  $1 \times 128079$  vector
  - "Unraveling" the matrix

- Note: All of these can be recast as the matrix that forms the image
  - Representations 2 and 4 are equivalent
    - The position is not represented

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## Example of a vector

- Vectors usually hold sets of numerical attributes

- $X, Y$ , value
  - $[1, 2, 0]$
- Earnings, losses, suicides
  - $[\$0 \ \$1,000,000 \ 3]$
- Etc ...

- Consider a "relative Manhattan" vector

- Provides a relative position by giving a number of avenues and streets to cross, e.g.  $[3\text{av } 33\text{st}]$



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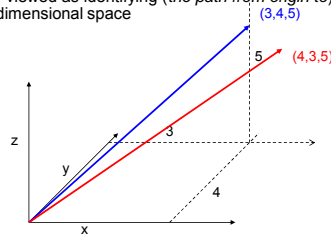
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## Vectors

- Ordered collection of numbers

- Examples:  $[3 \ 4 \ 5]$ ,  $[a \ b \ c \ d]$ , ...
- $[3 \ 4 \ 5] \neq [4 \ 3 \ 5]$  → Order is important

- Typically viewed as identifying (the path from origin to) a location in an  $N$ -dimensional space

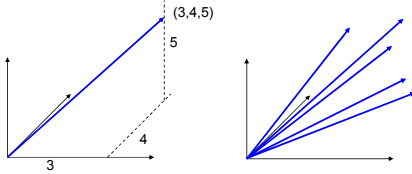


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## Vectors vs. Matrices



- A vector is a geometric notation for how to get from (0,0) to some location in the space
- A matrix is simply a collection of destinations!
  - Properties of matrices are *average* properties of the traveller's path to these destinations

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## Basic arithmetic operations

- Addition and subtraction
- Element-wise operations

$$\mathbf{a} + \mathbf{b} = \begin{bmatrix} a_1 \\ a_2 \\ a_3 \end{bmatrix} + \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix} = \begin{bmatrix} a_1 + b_1 \\ a_2 + b_2 \\ a_3 + b_3 \end{bmatrix} \quad \mathbf{a} - \mathbf{b} = \begin{bmatrix} a_1 \\ a_2 \\ a_3 \end{bmatrix} - \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix} = \begin{bmatrix} a_1 - b_1 \\ a_2 - b_2 \\ a_3 - b_3 \end{bmatrix}$$

$$\mathbf{A} + \mathbf{B} = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} + \begin{bmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{bmatrix} = \begin{bmatrix} a_{11} + b_{11} & a_{12} + b_{12} \\ a_{21} + b_{21} & a_{22} + b_{22} \end{bmatrix}$$

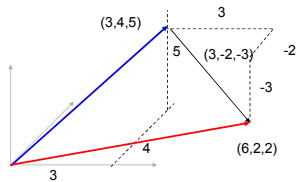
- MATLAB syntax: `a+b` and `a-b`

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## Vector Operations



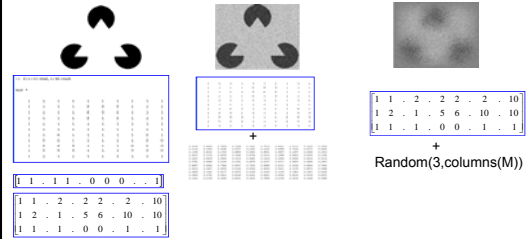
- Operations tell us how to get from  $\{0\}$  to the result of the vector operations
- $(3,4,5) + (3,-2,-3) = (6,2,2)$

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## Operations example



- Adding random values to different representations of the image

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## Vector norm

- Measure of how big a vector is:
  - Notated as  $\|\mathbf{x}\|$
  - $\| [a \ b \ \dots] \| = \sqrt{a^2 + b^2 + \dots}$
- In Manhattan vectors a measure of distance
  - $\| [-2 \ 17] \| = 17.11$
  - $\| [-6 \ 10] \| = 11.66$
- MATLAB syntax: `norm(x)`

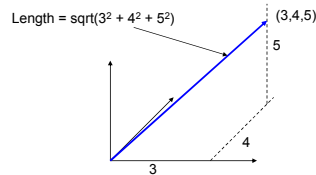


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## Vector Norm



- Geometrically the shortest distance to travel from the origin to the destination
  - As the crow flies
  - Assuming Euclidean Geometry

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## Transposition

- A transposed row vector becomes a column (and vice versa)

$$\mathbf{x} = \begin{bmatrix} a \\ b \\ c \end{bmatrix}, \mathbf{x}^T = [a \ b \ c] \quad \mathbf{y} = [a \ b \ c], \mathbf{y}^T = \begin{bmatrix} a \\ b \\ c \end{bmatrix}$$

- A transposed matrix gets all its row (or column) vectors transposed in order

$$\mathbf{X} = \begin{bmatrix} a & b & c \\ d & e & f \end{bmatrix}, \mathbf{X}^T = \begin{bmatrix} a & d \\ b & e \\ c & f \end{bmatrix} \quad \mathbf{M} = \begin{bmatrix} \text{img} \end{bmatrix}, \mathbf{M}^T = \begin{bmatrix} \text{img} \end{bmatrix}$$

- MATLAB syntax:  $\mathbf{a}'$

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## Vector multiplication

- Multiplication is not element-wise!
- Dot product, or inner product
  - Vectors must have the same number of elements
  - Row vector times column vector = **scalar**

$$[a \ b \ c] \cdot \begin{bmatrix} d \\ e \\ f \end{bmatrix} = a \cdot d + b \cdot e + c \cdot f$$

- Cross product, outer product or vector direct product
  - Column vector times row vector = **matrix**

$$\begin{bmatrix} a \\ b \\ c \end{bmatrix} \cdot [d \ e \ f] = \begin{bmatrix} a \cdot d & a \cdot e & a \cdot f \\ b \cdot d & b \cdot e & b \cdot f \\ c \cdot d & c \cdot e & c \cdot f \end{bmatrix}$$

- MATLAB syntax:  $\mathbf{a} * \mathbf{b}$

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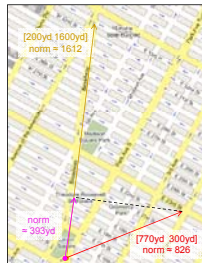
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## Vector *dot product* in Manhattan

- Multiplying the "yard" vectors
  - Instead of avenue/street we'll use yards
  - $\mathbf{a} = [200 \ 1600]$ ,  $\mathbf{b} = [770 \ 300]$
- The dot product of the two vectors relates to the length of a *projection*
  - How much of the first vector are we covered by following the second one?
  - The answer comes back as a unit of the first vector so we divide by its length

$$\frac{\mathbf{a} \cdot \mathbf{b}}{\|\mathbf{a}\|} = \frac{[200 \ 1600] \cdot [770 \ 300]}{\|[200 \ 1600]\|} \approx 393 \text{yd}$$

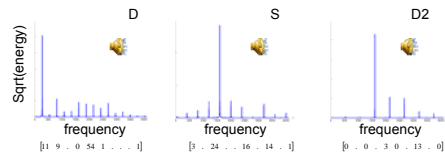


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## Vector dot product



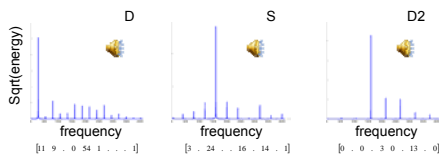
- Vectors are spectra
  - Energy at a discrete set of frequencies
  - Actually  $1 \times 4096$
  - X axis is the *index* of the number in the vector
    - Represents frequency
  - Y axis is the value of the number in the vector
    - Represents magnitude

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## Vector dot product



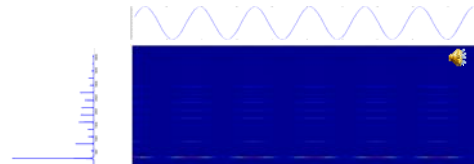
- How much of D is also in S
  - How much can you fake a D by playing an S
  - $D \cdot S / \|D\| \|S\| = 0.1$
  - Not very much
- How much of D is in D2?
  - $D \cdot D2 / \|D\| \|D2\| = 0.5$
  - Not bad, you can fake it
- To do this, D, S, and D2 *must be the same size*

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## Vector cross product



- The column vector is the spectrum
- The row vector is an amplitude modulation
- The crossproduct is a spectrogram
  - Shows how the energy in each frequency varies with time
  - The pattern in each column is a scaled version of the spectrum
  - Each row is a scaled version of the modulation

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## Matrix multiplication

- Generalization of vector multiplication

- Dot product of each vector pair

$$A \cdot B = \begin{bmatrix} \leftarrow & a_1 & \rightarrow \\ \leftarrow & a_2 & \rightarrow \end{bmatrix} \cdot \begin{bmatrix} \uparrow & b_1 & \downarrow \\ \uparrow & b_2 & \downarrow \end{bmatrix} = \begin{bmatrix} a_1 \cdot b_1 & a_1 \cdot b_2 \\ a_2 \cdot b_1 & a_2 \cdot b_2 \end{bmatrix}$$

- Dimensions must match!!
  - Columns of first matrix = rows of second
  - Result inherits the number of rows from the first matrix and the number of columns from the second matrix
- MATLAB syntax: `a * b`

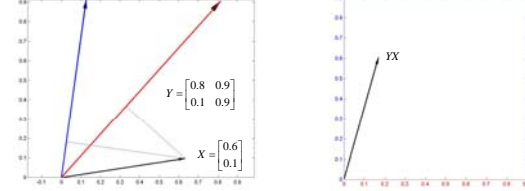
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## Multiplying a Vector by a Matrix

$$Y(2,:) = [0.1 \ 0.9] \quad Y(1,:) = [0.8 \ 0.9]$$



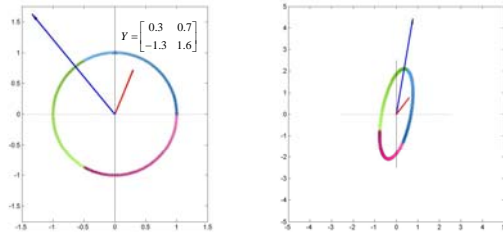
- Multiplication of a vector X by a matrix Y expresses the vector X in terms of projections of X on the row vectors of the matrix Y
  - It scales and rotates the vector
  - Alternately viewed, it scales and rotates the space – the underlying plane

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## Matrix Multiplication



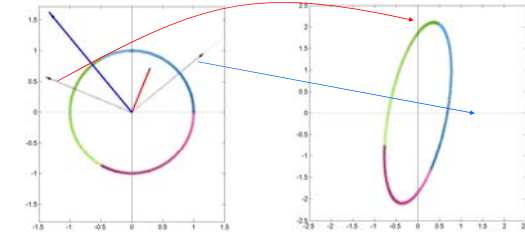
- The matrix rotates and scales the space
  - Including its own vectors

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## Matrix Multiplication



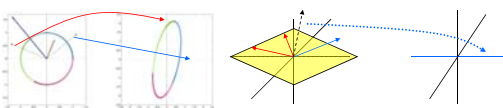
- The *normals* to the row vectors in the matrix become the new axes
  - X axis = normal to the *second* row vector
    - Scaled by the inverse of the length of the *first* row vector

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## Matrix Multiplication is projection



- The k-th axis corresponds to the normal to the hyperplane represented by the 1..k-1, k+1..N-th row vectors in the matrix
  - Any set of K-1 vectors represent a hyperplane of dimension K-1 or less
- The distance along the new axis equals the length of the projection on the k-th row vector
  - Expressed in inverse-lengths of the vector

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## Matrix Multiplication: Column space

$$\begin{bmatrix} a & b & c \\ d & e & f \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} = x \begin{bmatrix} a \\ d \end{bmatrix} + y \begin{bmatrix} b \\ e \end{bmatrix} + z \begin{bmatrix} c \\ f \end{bmatrix}$$

- So much for spaces .. what does multiplying a matrix by a vector really do?
- It *mixes* the column vectors of the matrix using the numbers in the vector
- The *column space* of the Matrix is the complete set of all vectors that can be formed by mixing its columns

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## Matrix Multiplication: Row space

$$\begin{bmatrix} x & y \end{bmatrix} \begin{bmatrix} a & b & c \\ d & e & f \end{bmatrix} = x \begin{bmatrix} a & b & c \end{bmatrix} + y \begin{bmatrix} d & e & f \end{bmatrix}$$

- Left multiplication mixes the *row* vectors of the matrix.
- The *row space* of the Matrix is the complete set of all vectors that can be formed by mixing its rows

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## Matrix multiplication: Mixing vectors

$$\begin{bmatrix} 1 & 3 & 0 \\ \cdot & \cdot & 0 \\ 9 & 24 & \cdot \\ \cdot & \cdot & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} = \begin{bmatrix} 7 \\ \cdot \\ \cdot \\ 2 \end{bmatrix}$$

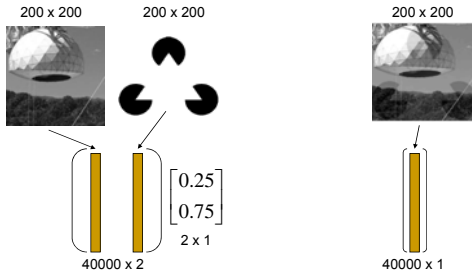
- A physical example
  - The three column vectors of the matrix X are the spectra of three notes
  - The multiplying column vector Y is just a mixing vector
  - The result is a sound that is the mixture of the three notes

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## Matrix multiplication: Mixing vectors



- Mixing two images
  - The images are arranged as columns
  - position value not included
  - The result of the multiplication is rearranged as an image

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## Matrix multiplication: another view

$$\mathbf{A} \cdot \mathbf{B} = \begin{bmatrix} a_{11} & \dots & a_{1N} \\ a_{21} & \dots & a_{2N} \\ \dots & \dots & \dots \\ a_{M1} & \dots & a_{MN} \end{bmatrix} \begin{bmatrix} b_{11} & \dots & b_{1K} \\ \dots & \dots & \dots \\ b_{N1} & \dots & b_{NK} \end{bmatrix} = \begin{bmatrix} \sum_k a_{1k} b_{k1} & \dots & \sum_k a_{1k} b_{kK} \\ \dots & \dots & \dots \\ \sum_k a_{Mk} b_{k1} & \dots & \sum_k a_{Mk} b_{kK} \end{bmatrix}$$

- What does this mean?

$$\begin{bmatrix} a_{11} & \dots & a_{1N} \\ a_{21} & \dots & a_{2N} \\ \dots & \dots & \dots \\ a_{M1} & \dots & a_{MN} \end{bmatrix} \begin{bmatrix} b_{11} & \dots & b_{1K} \\ \dots & \dots & \dots \\ b_{N1} & \dots & b_{NK} \end{bmatrix} = \begin{bmatrix} a_{11} \\ \dots \\ a_{M1} \end{bmatrix} \begin{bmatrix} b_{11} & \dots & b_{1K} \end{bmatrix} + \begin{bmatrix} a_{12} \\ \dots \\ a_{M2} \end{bmatrix} \begin{bmatrix} b_{21} & \dots & b_{2K} \end{bmatrix} + \dots + \begin{bmatrix} a_{1N} \\ \dots \\ a_{MN} \end{bmatrix} \begin{bmatrix} b_{N1} & \dots & b_{NK} \end{bmatrix}$$

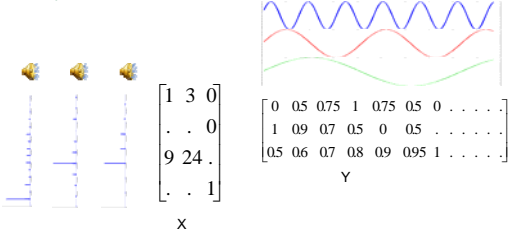
- The outer product of the first column of A and the first row of B + outer product of the second column of A and the second row of B + ....

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## Why is that useful?



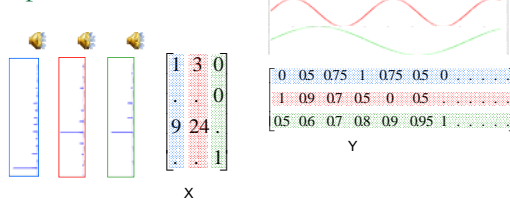
- Sounds: Three notes modulated independently

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## Matrix multiplication: Mixing modulated spectra



- Sounds: Three notes modulated independently

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### Matrix multiplication: Mixing modulated spectra

$$\begin{bmatrix} 1 \\ 9 \end{bmatrix} \times \begin{bmatrix} 0 & 0.5 & 0.75 & 1 & 0.75 & 0.5 & 0 & \dots & \dots & \dots \end{bmatrix}$$

- Sounds: Three notes modulated independently

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### Matrix multiplication: Mixing modulated spectra

$$\begin{bmatrix} 3 \\ 24 \end{bmatrix} \times \begin{bmatrix} 1 & 0.9 & 0.7 & 0.5 & 0 & 0.5 & \dots & \dots & \dots & \dots \end{bmatrix}$$

- Sounds: Three notes modulated independently

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### Matrix multiplication: Mixing modulated spectra

$$\begin{bmatrix} 0 \\ 0 \end{bmatrix} \times \begin{bmatrix} 0.5 & 0.6 & 0.7 & 0.8 & 0.9 & 0.95 & 1 & \dots & \dots & \dots \end{bmatrix}$$

- Sounds: Three notes modulated independently

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### Matrix multiplication: Mixing modulated spectra

$$\begin{bmatrix} \dots \\ \dots \\ \dots \end{bmatrix} \times \begin{bmatrix} \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \end{bmatrix}$$

- Sounds: Three notes modulated independently

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### Matrix multiplication: Image transition

$$\begin{bmatrix} i_1 \\ i_2 \end{bmatrix} \times \begin{bmatrix} 1 & 0.9 & 0.8 & 0.7 & 0.6 & 0.5 & 0.4 & 0.3 & 0.2 & 0.1 \\ 0 & 1 & 2 & 3 & 4 & 5 & 6 & 7 & 8 & 9 \end{bmatrix}$$

- Image1 fades out linearly
- Image 2 fades in linearly

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### Matrix multiplication: Image transition

$$\begin{bmatrix} i_1 \\ i_2 \\ \dots \\ i_N \end{bmatrix} \times \begin{bmatrix} 1 & 0.9i_1 & 0.8i_1 & \dots & 0 \\ 0 & 0.9i_2 & 0.8i_2 & \dots & 0 \\ \dots & \dots & \dots & \dots & \dots \\ 0 & 0.9i_N & 0.8i_N & \dots & 0 \end{bmatrix}$$

- Each column is one image
  - The columns represent a sequence of images of decreasing intensity
- Image1 fades out linearly

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### Matrix multiplication: Image transition

- Image 2 fades in linearly

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### Matrix multiplication: Image transition

- Image 1 fades out linearly
- Image 2 fades in linearly

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### The Identity Matrix

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

- An identity matrix is a square matrix where
  - All diagonal elements are 1.0
  - All off-diagonal elements are 0.0
- Multiplication by an identity matrix does not change vectors

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### Diagonal Matrix

$$y = \begin{bmatrix} 2 & 0 \\ 0 & 1 \end{bmatrix}$$

- All off-diagonal elements are zero
- Diagonal elements are non-zero
- Scales the axes
  - May flip axes

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### Diagonal matrix to transform images

- How?

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### Stretching

$$\begin{bmatrix} 2 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 2 & 2 & 2 & 2 & 10 \\ 1 & 2 & 1 & 5 & 6 & 10 & 10 \\ 1 & 1 & 1 & 0 & 0 & 1 & 1 \end{bmatrix}$$

- Location-based representation
- Scaling matrix – only scales the X axis
  - The Y axis and pixel value are scaled by identity
- Not a good way of scaling.

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### Stretching

$D =$

$$A = \begin{bmatrix} 1 & .5 & 0 & 0 \\ 0 & .5 & 1 & .5 \\ 0 & 0 & 0 & .5 \\ 0 & 0 & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots \end{bmatrix} \quad (Nx2N)$$

Newpic = EA

- Better way

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### Modifying color

$P =$

$$Newpic = P \begin{bmatrix} R & G & B \\ 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

- Scale only Green

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### Permutation Matrix

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

- A permutation matrix simply rearranges the axes
  - The row entries are axis vectors in a different order
  - The result is a combination of rotations and reflections
- The permutation matrix effectively *permutes* the arrangement of the elements in a vector

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### Permutation Matrix

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

$\begin{bmatrix} 1 & 1 & 2 & 2 & 2 & 2 & 10 \\ 1 & 2 & 1 & 5 & 6 & 10 & 10 \\ 1 & 1 & 1 & 0 & 0 & 1 & 1 \end{bmatrix}$

- Reflections and 90 degree rotations of images and objects

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### Permutation Matrix

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

$\begin{bmatrix} x_1 & x_2 & \dots & x_N \\ y_1 & y_2 & \dots & y_N \\ z_1 & z_2 & \dots & z_N \end{bmatrix}$

- Reflections and 90 degree rotations of images and objects
  - Object represented as a matrix of 3-Dimensional "position" vectors
  - Positions identify each point on the surface

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### Rotation Matrix

$x' = x \cos \theta - y \sin \theta$   
 $y' = x \sin \theta + y \cos \theta$

$R_\theta = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$

$R_\theta X = X_{new}$

$X = \begin{bmatrix} x \\ y \end{bmatrix}$   
 $X_{new} = \begin{bmatrix} x' \\ y' \end{bmatrix}$

- A rotation matrix *rotates* the vector by some angle  $\theta$
- Alternately viewed, it rotates the axes
  - The new axes are at an angle  $\theta$  to the old one

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### Rotating a picture

$$R = \begin{bmatrix} \cos 45 & -\sin 45 & 0 \\ \sin 45 & \cos 45 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 1 & 2 & 2 & 2 & 2 & 2 & 2 & 2 & 2 \\ 1 & 2 & 1 & 5 & 6 & 10 & 10 & 10 & 10 & 10 \\ 1 & 1 & 1 & 0 & 0 & 1 & 1 & 1 & 1 & 1 \end{bmatrix} \rightarrow \begin{bmatrix} 0 & -\sqrt{2} & \sqrt{2} & -3\sqrt{2} & -4\sqrt{2} & -8\sqrt{2} & -8\sqrt{2} & -8\sqrt{2} & -8\sqrt{2} & -8\sqrt{2} \\ \sqrt{2} & 3\sqrt{2} & 3\sqrt{2} & 7\sqrt{2} & 8\sqrt{2} & 12\sqrt{2} & 12\sqrt{2} & 12\sqrt{2} & 12\sqrt{2} & 12\sqrt{2} \\ 1 & 1 & 1 & 0 & 0 & 1 & 1 & 1 & 1 & 1 \end{bmatrix}$$

- Note the representation: 3-row matrix
  - Rotation only applies on the "coordinate" rows
  - The value does not change
  - Why is pacman grainy?

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### 3-D Rotation

- 2 degrees of freedom
  - 2 separate angles
- What will the rotation matrix be?

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### Projections

- What would we see if the cone to the left were transparent if we looked at it along the normal to the plane
  - The plane goes through the origin
  - Answer: the figure to the right
- How do we get this? Projection

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### Projection Matrix

- Consider any plane specified by a set of vectors  $W_1, W_2, \dots$ 
  - Or matrix  $[W_1, W_2, \dots]$
  - Any vector can be projected onto this plane
  - The matrix  $A$  that rotates and scales the vector so that it becomes its projection is a projection matrix

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### Projection Matrix

- Given a set of vectors  $W_1, W_2$ , which form a matrix  $W = [W_1, W_2, \dots]$
- The projection matrix that transforms any vector  $X$  to its projection on the plane is
  - $P = W(W^T W)^{-1} W^T$ 
    - We will visit matrix inversion shortly
- Magic – any set of vectors from the same plane that are expressed as a matrix will give you the same projection matrix
  - $P = V(V^T V)^{-1} V^T$

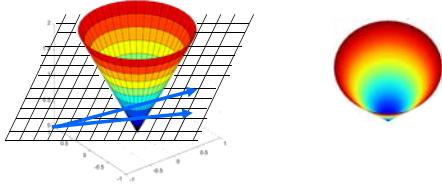
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### Projections

- HOW?

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## Projections



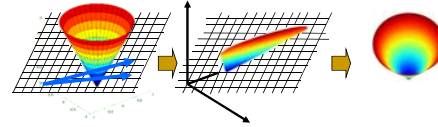
- Draw any two vectors  $W_1$  and  $W_2$  that lie on the plane
  - **ANY two so long as they have different angles**
- Compose a matrix  $W = [W_1 \ W_2]$
- Compose the projection matrix  $P = W (W^T W)^{-1} W^T$
- Multiply every point on the cone by  $P$  to get its projection
- View it ☺
  - I'm missing a step here – what is it?

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## Projections



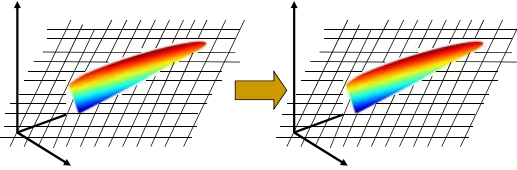
- The projection actually projects it onto the plane, but you're still seeing the plane in 3D
  - The result of the projection is a 3-D vector
  - $P = W (W^T W)^{-1} W^T = 3 \times 3$ ,  $P \cdot \text{Vector} = 3 \times 1$
  - The image must be rotated till the plane is in the plane of the paper
    - The Z axis in this case will always be zero and can be ignored
    - How will you rotate it? (remember you know  $W_1$  and  $W_2$ )

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## Projection matrix properties



- The projection of any vector that is already on the plane is the vector itself
  - $Px = x$  if  $x$  is on the plane
  - If the object is already on the plane, there is no further projection to be performed
- The projection of a projection is the projection
  - $P(Px) = Px$
  - That is because  $Px$  is already on the plane
- Projection matrices are *idempotent*
  - $P^2 = P$
  - Follows from the above

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