#### Week 1 Review Session

#### Gov January Linear Algebra Review Soubhik Barari Harvard University

Jan 11, 2021

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Why should we care? *Every* statistical method in social science explicitly or implicitly relies on this accounting system.

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- Summarising and combining vectors
  - Length (Schwarz inequality, Triangle inequality)
  - Dot products
- Gaussian elimination
- Gauss-Jordan elimination
- Matrix inversion

#### Plan

Cover four problems  $\leadsto$  discuss intuition and connect them to a real application in political science.

- ▶ Problem 3 (1.2.2): summarising, combining vectors
- ▶ Problem 6 (2.1.9): transforming matrices
- ▶ Problem 9 (2.2.1-2): solving unknowns between matrices
- ▶ Problem 11 (2.5.12): inverting matrices

Don't feel self-conscious about interrupting if you're confused or have questions!

The simplest linear algebraic unit is a **vector** in some space (e.g.  $\mathbb{R}^2$ ), where each dimension usually "means something":

$$u = \begin{bmatrix} -0.6 \\ 0.8 \end{bmatrix}, v = \begin{bmatrix} 4 \\ 3 \end{bmatrix}, w = \begin{bmatrix} 1 \\ 2 \end{bmatrix} \leftarrow$$
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We can summarise a vector by its length (or norm):

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But, how does this dot product relate to their individual lengths?

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Indeed, we can confirm this is true for our vectors!

$$|v \cdot w| = |4 \cdot 1 + 3 \cdot 2| = 10 < 5\sqrt{5} = ||v|| \cdot ||w||$$

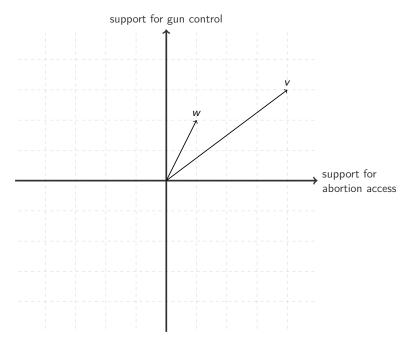
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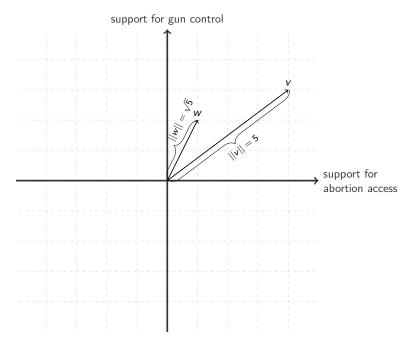
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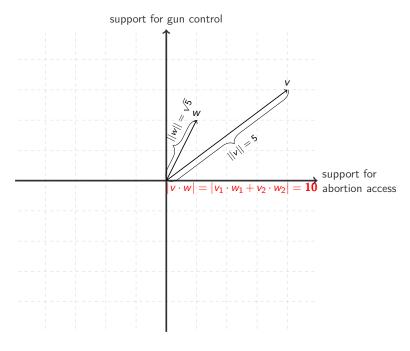
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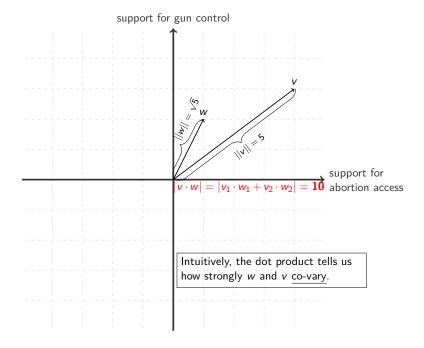
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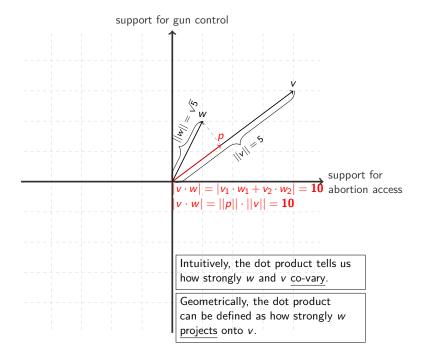
Let's visualize this in our example space.

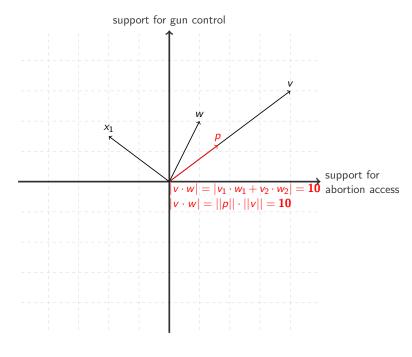


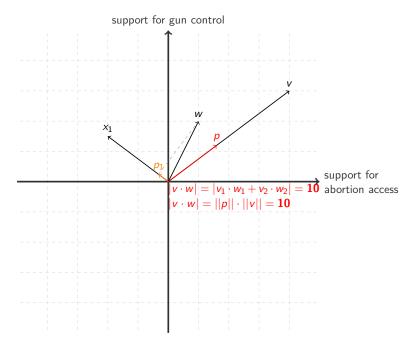


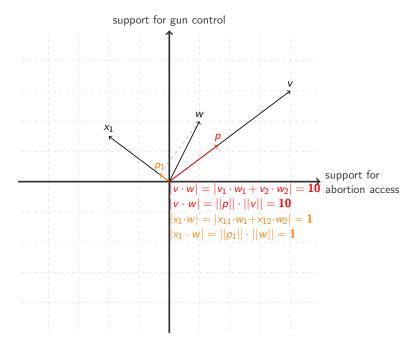


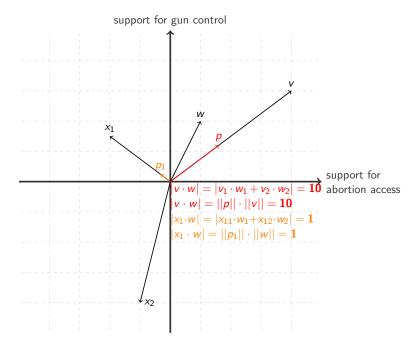


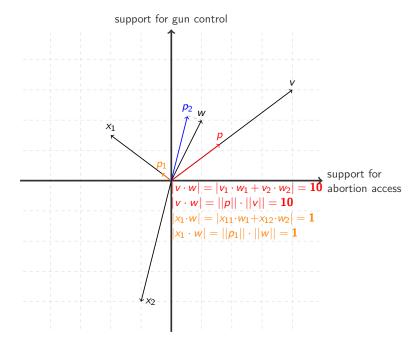


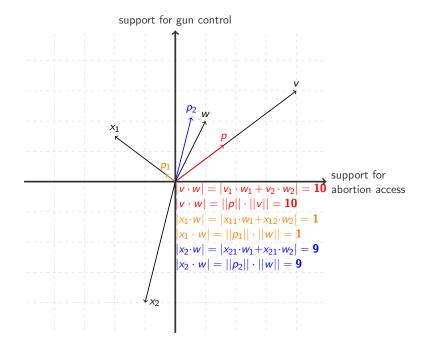












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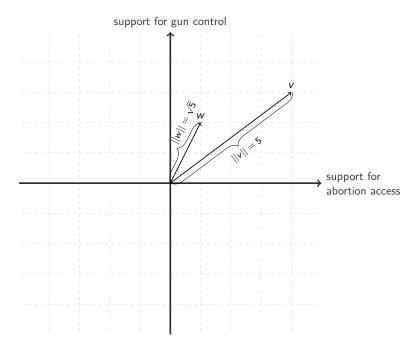
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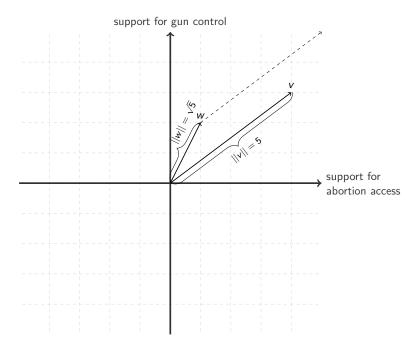
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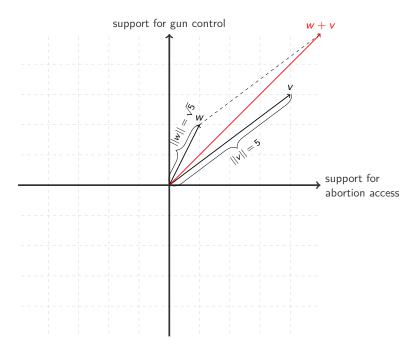
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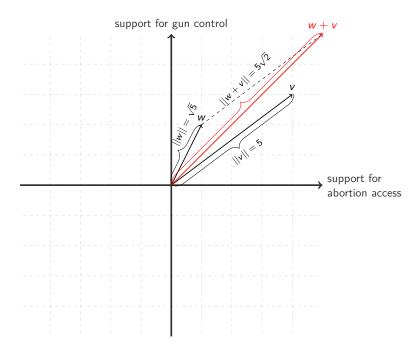
In our case:

$$||v + w|| = 5\sqrt{2} \le 5 + \sqrt{5} = ||v|| + ||w||$$









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	free trade	abortion access	gun control	\
Maria {	1	2	4	
Dev{	-2	3	1	= A
Jinyang{	_4	1	2	]

Suppose we want to scale down each person's views to a single ideology measure in  $\mathbb{R}$  and collect it in a vector b. How would we do this?

Let's say we have used some algorithms to compute a vector of the U.S. Democratic party platform's intensity of support/opposition of these three issues  $x = [1, 3, 4]^T$ .

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One method for ideologically scaling our voters is treating their positions as a **linear transformation** of the party's positions  $\rightsquigarrow$  **A**x.

$$\mathbf{A}x = \begin{vmatrix} 1 & 2 & 4 \\ -2 & 3 & 1 \\ -4 & 1 & 2 \end{vmatrix} \begin{vmatrix} 1 \\ 3 \\ 4 \end{vmatrix} = b$$

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There are two ways to conduct a linear transformation.

$$\mathbf{A}x = \begin{bmatrix} 1 & 2 & 4 \\ -2 & 3 & 1 \\ -4 & 1 & 2 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 3 \\ 4 \end{bmatrix} = \begin{bmatrix} ? \\ ? \\ ? \end{bmatrix}$$

$$\mathbf{A}x = \begin{bmatrix} 1 & 2 & 4 \\ -2 & 3 & 1 \\ -4 & 1 & 2 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 3 \\ 4 \end{bmatrix} = \begin{bmatrix} 23 \\ ? \\ ? \end{bmatrix}$$

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1. As a dot product of row vectors with x:

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2. As a sum of column vectors weighted by x:

$$\mathbf{A}x = \begin{bmatrix} 1 & 2 & 4 \\ -2 & 3 & 1 \\ -4 & 1 & 2 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 3 \\ 4 \end{bmatrix} = 1 \cdot \begin{bmatrix} 1 \\ -2 \\ 4 \end{bmatrix} + 3 \cdot \begin{bmatrix} 2 \\ 3 \\ 1 \end{bmatrix} + 4 \cdot \begin{bmatrix} 4 \\ 1 \\ 2 \end{bmatrix} = \begin{bmatrix} 23 \\ 11 \\ 7 \end{bmatrix}$$

In either case, the results b of the linear transformation of the party positions (x) by our voters' issue positions (A) reveals Maria to be the "strongest Democrat".

Now, suppose that we've collected two voters' issue positions,  $\mathbf{A}$ , and we already asked them to numerically scale the strength of their Republican-Democrat affiliation (in  $\mathbb{R}$ ), collected as vector  $\mathbf{b}$ .

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This time, we don't have measures of x, the "Democratic-ness" of support for each issue. Can we solve for this, though?

$$\mathbf{A}x = b \quad \sim \quad \begin{bmatrix} 2 & 3 \\ 4 & 1 \end{bmatrix} \cdot \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 17 \end{bmatrix}$$

**Gaussian elimination**: A generalized "linear algebraic" way of solving systems of equations.

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<u>Step 2.</u> Reduce augmented matrix, if possible, until an upper triangular matrix appears on left-hand side (row echelon form),

$$\left[\begin{array}{cc|c} U_{11} & U_{12} & b_1' \\ 0 & U_{22} & b_2' \end{array}\right]$$

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4-4 & 1-6 & 17-2
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$$\begin{bmatrix} 2 & 3 & 1 \\ 0 & -5 & 15 \end{bmatrix} \sim 2x_1 = 10 \\ x_2 = -3$$

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Option a: In system of equations, start with most obvious solution, and keep substituting in to get others (back-substitution).

$$\begin{bmatrix} 2 & 3 & 1 \\ 0 & -5 & 15 \end{bmatrix} \sim x_1 = \frac{5}{x_2} = -3$$

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$$\begin{bmatrix} 2 & 3 & 1 \\ 0 & -5 & 15 \end{bmatrix} \cdot -\frac{1}{5}$$

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$$\begin{bmatrix} 2 & 3 & 1 \\ 0 & 1 & -3 \end{bmatrix} + (-3 \cdot \text{row } 2)$$

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$$\left[\begin{array}{c|c} 2 & 0 & 10 \\ 0 & 1 & -3 \end{array}\right] \cdot \frac{1/2}{2}$$

Step 4. Extract solution for x, if it exists.

*Option a*: In system of equations, start with most obvious solution, and keep substituting in to get others (back-substitution).

$$\begin{bmatrix} 2 & 3 & 1 \\ 0 & -5 & 15 \end{bmatrix} \sim x_1 = \mathbf{5}$$
$$x_2 = -\mathbf{3}$$

$$\left[\begin{array}{cc|c} 1 & 0 & \mathbf{5} \\ 0 & 1 & -\mathbf{3} \end{array}\right]$$

Rashida 
$$\left\{ \begin{array}{ccc} & & & & \\ & 2 & & & \\ & & 4 & & 1 \end{array} \right) \cdot \begin{bmatrix} \mathbf{5} \\ -\mathbf{3} \end{bmatrix} = \begin{bmatrix} 1 \\ 17 \end{bmatrix}$$

$$\text{Rashida} \left\{ \left( \begin{array}{ccc} \textbf{2} & \textbf{3} \\ \textbf{3} & \textbf{1} \end{array} \right) \cdot \begin{bmatrix} \textbf{5} \\ \textbf{-3} \end{bmatrix} = \begin{bmatrix} 1 \\ 17 \end{bmatrix}$$

In either case, the solution x tells us that:

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In either case, the solution x tells us that:

- ▶ support for public healthcare is a Democratic position → increased support = increases Democratic affiliation by 5
- ▶ support for military spending is a Republican position  $\rightsquigarrow$  increased support = decreases Democratic affiliation by -3

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In many cases, there are not enough "independent" columns or rows in  $\mathbf{A} \rightsquigarrow$  we fail  $\underline{\mathsf{Step 2}}$  and cannot find exactly one solution (more this week).

Another method for solving  $\mathbf{A}x = b$  would be eliminating  $\mathbf{A}$  from the left-hand side of the equation completely via its **inverse**:

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$$\label{eq:constraints} \begin{split} & \textbf{C} & = \textbf{A}\textbf{B} \\ \Longrightarrow \textbf{A}^{-1}\textbf{C} & = \textbf{A}^{-1}\textbf{A}\textbf{B} \\ \Longrightarrow \textbf{A}^{-1}\textbf{C}\textbf{C}^{-1} & = \textbf{B}\textbf{C}^{-1} \end{split}$$

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Ok, so how exactly do you invert a non-singular matrix?

All we have to do is to solve:

$$\begin{bmatrix} 2 & 3 \\ 4 & 1 \end{bmatrix} \begin{bmatrix} a & b \\ c & d \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

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$$\begin{bmatrix} 2 & 3 \\ 4 & 1 \end{bmatrix} \begin{bmatrix} a \\ c \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \qquad \begin{bmatrix} 2 & 3 \\ 4 & 1 \end{bmatrix} \begin{bmatrix} b \\ d \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

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$$\begin{bmatrix} 2 & 3 & 1 \\ 4 & 1 & 0 \end{bmatrix}, \qquad \begin{bmatrix} 2 & 3 & 0 \\ 4 & 1 & 1 \end{bmatrix}$$

$$\left[\begin{array}{cc|cccc}
2 & 3 & 1 & 0 \\
0 & -5 & -2 & 1
\end{array}\right]$$

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$$\begin{bmatrix} 2 & 3 & 1 \\ 4 & 1 & 0 \end{bmatrix}, \qquad \begin{bmatrix} 2 & 3 & 0 \\ 4 & 1 & 1 \end{bmatrix}$$

$$\left[\begin{array}{c|cc|c}
2 & 3 & 1 & 0 \\
0 & 1 & 2/5 & -1/5
\end{array}\right]$$

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$$\begin{bmatrix} 2 & 3-3 & | & 1-6/5 & 0+3/5 \\ 0 & 1 & | & 2/5 & -1/5 \end{bmatrix} + (-3 \cdot \text{row 2})$$

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$$\left|\begin{array}{cc|c} 1 & 0 & -1/10 & 3/10 \\ 0 & 1 & 2/5 & -1/5 \end{array}\right| \cdot \frac{1}{2}$$

All we have to do is to solve:

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$$\begin{bmatrix} 2 & 3 & 1 \\ 4 & 1 & 0 \end{bmatrix}, \qquad \begin{bmatrix} 2 & 3 & 0 \\ 4 & 1 & 1 \end{bmatrix}$$

$$\left[\begin{array}{cc|cc} 1 & 0 & -0.1 & 0.3 \\ 0 & 1 & 0.4 & -0.2 \end{array}\right]$$

```
We can verify this in R:
```

```
A <- matrix(c(2,4,3,1), ncol=2, nrow=2) solve(A)
```

```
## [,1] [,2]
## [1,] -0.1 0.3
## [2,] 0.4 -0.2
```

#### This Week

#### Suggested concepts to focus on:

- Space and subspace
  - column space
  - row space
  - nullspace
- Linear independence
  - rank
  - basis

#### Questions?