

Activity

Suppose you have available a procedure `is_independent(L)`, which takes a list `L` of `Vecs` and returns `True` or `False` depending on whether the vectors are independent or not.

Write a procedure

```
def has_many_solutions(a_list, b_list, u)...
```

with the following spec:

- ▶ **input:** list $[\mathbf{a}_1, \dots, \mathbf{a}_n]$ of `Vecs`, list $[\beta_1, \dots, \beta_n]$ of scalars, `Vec` \mathbf{u} that is a solution to the linear system

$$\begin{aligned}\mathbf{a}_1 \cdot \mathbf{x} &= \beta_1 \\ &\vdots \\ \mathbf{a}_n \cdot \mathbf{x} &= \beta_n\end{aligned}$$

- ▶ **output:** `True` if there are solutions other than \mathbf{u} to the linear system.

Activity

So far we've done *paths = spanning* and *cycles = linearly dependent* over $\text{GF}(2)$. How would you achieve the same over \mathbb{R} ?

Properties of linear (in)dependence

Linear-Dependence Lemma Let $\mathbf{v}_1, \dots, \mathbf{v}_n$ be vectors. A vector \mathbf{v}_i is in the span of the other vectors if and only if the zero vector can be written as a linear combination of $\mathbf{v}_1, \dots, \mathbf{v}_n$ in which the coefficient of \mathbf{v}_i is nonzero.

Contrapositive:

\mathbf{v}_i is *not* in the space of the other vectors
if and only if
for any linear combination equaling the zero vector
$$\mathbf{0} = \alpha_1 \mathbf{v}_1 + \dots + \alpha_i \mathbf{v}_i + \dots + \alpha_n \mathbf{v}_n$$

it must be that the coefficient α_i is zero.

Analyzing the Grow algorithm

```
def GROW( $\mathcal{V}$ )  
   $S = \emptyset$   
  repeat while possible:  
    find a vector  $\mathbf{v}$  in  $\mathcal{V}$  that is not in  $\text{Span } S$ , and put it in  $S$ .
```

Grow-Algorithm Corollary: The vectors obtained by the Grow algorithm are linearly independent.

In graphs, this means that the solution obtained by the Grow algorithm has no cycles (is a forest).

Analyzing the Grow algorithm

Grow-Algorithm Corollary: The vectors obtained by the Grow algorithm are linearly independent.

Proof: For $n = 1, 2, \dots$, let \mathbf{v}_n be the vector added to S in the n^{th} iteration of the Grow algorithm. We show by induction that $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$ are linearly independent.

For $n = 0$, there are no vectors, so the claim is trivially true.

Assume the claim is true for $n = k - 1$. We prove it for $n = k$.

The vector \mathbf{v}_k added to S in the k^{th} iteration is not in the span of $\mathbf{v}_1, \dots, \mathbf{v}_{k-1}$.

Therefore, by the Linear-Dependence Lemma, for any coefficients $\alpha_1, \dots, \alpha_k$ such that

$$\mathbf{0} = \alpha_1 \mathbf{v}_1 + \dots + \alpha_{k-1} \mathbf{v}_{k-1} + \alpha_k \mathbf{v}_k$$

it must be that α_k equals zero. We may therefore write

$$\mathbf{0} = \alpha_1 \mathbf{v}_1 + \dots + \alpha_{k-1} \mathbf{v}_{k-1}$$

By claim for $n = k - 1$, $\mathbf{v}_1, \dots, \mathbf{v}_{k-1}$ are linearly independent, so $\alpha_1 = \dots = \alpha_{k-1} = 0$

The linear combination of $\mathbf{v}_1, \dots, \mathbf{v}_k$ is *trivial*. We have proved that $\mathbf{v}_1, \dots, \mathbf{v}_k$ are linearly independent. This proves the claim for $n = k$. QED

Analyzing the Shrink algorithm

def SHRINK(\mathcal{V})

S = some finite set of vectors that spans \mathcal{V}

repeat while possible:

find a vector \mathbf{v} in S such that $\text{Span}(S - \{\mathbf{v}\}) = \mathcal{V}$, and remove \mathbf{v} from S .

Shrink-Algorithm Corollary: The vectors obtained by the Shrink algorithm are linearly independent.

In graphs, this means that the Shrink algorithm outputs a solution that is a forest.

Recall:

Superfluous-Vector Lemma For any set S and any vector $\mathbf{v} \in S$, if \mathbf{v} can be written as a linear combination of the other vectors in S then

$$\text{Span}(S - \{\mathbf{v}\}) = \text{Span } S$$

Analyzing the Shrink algorithm

Shrink-Algorithm Corollary: The vectors obtained by the Shrink algorithm are linearly independent.

Proof: Let $S = \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$ be the set of vectors obtained by the Shrink algorithm. Assume for a contradiction that the vectors are linearly dependent.

Then $\mathbf{0}$ can be written as a nontrivial linear combination

$$\mathbf{0} = \alpha_1 \mathbf{v}_1 + \dots + \alpha_n \mathbf{v}_n$$

where at least one of the coefficients is nonzero.

Let α_j be one of the nonzero coefficients.

By the Linear-Dependence Lemma, \mathbf{v}_j can be written as a linear combination of the other vectors.

Hence by the Superfluous-Vector Lemma, $\text{Span}(S - \{\mathbf{v}_j\}) = \text{Span } S$, so the Shrink algorithm should have removed \mathbf{v}_j .

QED

Basis

If they successfully finish, the Grow algorithm and the Shrink algorithm each find a set of vectors spanning the vector space \mathcal{V} . In each case, the set of vectors found is linearly independent.

Definition: Let \mathcal{V} be a vector space. A *basis* for \mathcal{V} is a linearly independent set of generators for \mathcal{V} .

Thus a set S of vectors of \mathcal{V} is a *basis* for \mathcal{V} if S satisfies two properties:

Property B1 (*Spanning*) $\text{Span } S = \mathcal{V}$, and

Property B2 (*Independent*) S is linearly independent.

Most important definition in linear algebra.

Basis: Examples

A set S of vectors of \mathcal{V} is a *basis* for \mathcal{V} if S satisfies two properties:

Property B1 (*Spanning*) $\text{Span } S = \mathcal{V}$, and

Property B2 (*Independent*) S is linearly independent.

Example: Let $\mathcal{V} = \text{Span} \{[1, 0, 2, 0], [0, -1, 0, -2], [2, 2, 4, 4]\}$.

Is $\{[1, 0, 2, 0], [0, -1, 0, -2], [2, 2, 4, 4]\}$ a basis for \mathcal{V} ?

The set *is* spanning but is *not* independent

$$1 [1, 0, 2, 0] - 1 [0, -1, 0, -2] - \frac{1}{2} [2, 2, 4, 4] = \mathbf{0}$$

so not a basis

However, $\{[1, 0, 2, 0], [0, -1, 0, -2]\}$ *is* a basis:

- ▶ Obvious that these vectors are independent because each has a nonzero entry where the other has a zero.
- ▶ To show

$\text{Span} \{[1, 0, 2, 0], [0, -1, 0, -2]\} = \text{Span} \{[1, 0, 2, 0], [0, -1, 0, -2], [2, 2, 4, 4]\}$,
can use Superfluous-Vector Lemma:

$$[2, 2, 4, 4] = 2 [1, 0, 2, 0] - 2 [0, -1, 0, -2]$$

Basis: Examples

Example: A simple basis for \mathbb{R}^3 : the standard generators $\mathbf{e}_1 = [1, 0, 0]$, $\mathbf{e}_2 = [0, 1, 0]$, $\mathbf{e}_3 = [0, 0, 1]$.

- ▶ *Spanning:* For any vector $[x, y, z] \in \mathbb{R}^3$,

$$[x, y, z] = x [1, 0, 0] + y [0, 1, 0] + z [0, 0, 1]$$

- ▶ *Independent:* Suppose

$$\mathbf{0} = \alpha_1 [1, 0, 0] + \alpha_2 [0, 1, 0] + \alpha_3 [0, 0, 1] = [\alpha_1, \alpha_2, \alpha_3]$$

Then $\alpha_1 = \alpha_2 = \alpha_3 = 0$.

Instead of “standard generators”, we call them *standard basis vectors*.

We refer to $\{[1, 0, 0], [0, 1, 0], [0, 0, 1]\}$ as *standard basis* for \mathbb{R}^3 .

In general the standard generators are usually called *standard basis vectors*.

Basis: Examples

Example: Another basis for \mathbb{R}^3 : $[1, 1, 1], [1, 1, 0], [0, 1, 1]$

- ▶ *Spanning:* Can write standard generators in terms of these vectors:

$$[1, 0, 0] = [1, 1, 1] - [0, 1, 1]$$

$$[0, 1, 0] = [1, 1, 0] + [0, 1, 1] - [1, 1, 1]$$

$$[0, 0, 1] = [1, 1, 1] - [1, 1, 0]$$

Since $\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3$ can be written in terms of these new vectors, every vector in $\text{Span}\{\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3\}$ is in span of new vectors.

Thus \mathbb{R}^3 equals span of new vectors.

- ▶ *Linearly independent:* Write zero vector as linear combination:

$$\mathbf{0} = x[1, 1, 1] + y[1, 1, 0] + z[0, 1, 1] = [x + y, x + y + z, x + z]$$

Looking at each entry, we get

$$0 = x + y$$

$$0 = x + y + z$$

$$0 = x + z$$

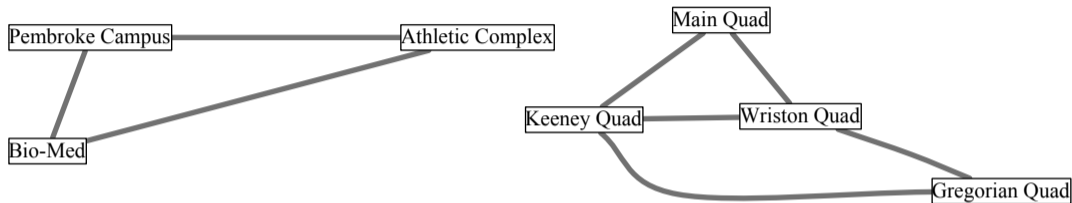
Plug $x + y = 0$ into second equation to get $0 = z$.

Plug $z = 0$ into third equation to get $x = 0$.

Plug $x = 0$ into first equation to get $y = 0$.

Thus the linear combination is trivial.

Basis: Examples in graphs



One kind of basis in a graph G : a set S of edges forming a spanning forest.

- ▶ *Spanning*: for each edge xy in G , there is an x -to- y path consisting of edges of S .
- ▶ *Independent*: no cycle consisting of edges of S

Towards showing that every vector space has a basis

We would like to prove that every vector space \mathcal{V} has a basis.

The Grow algorithm and the Shrink algorithm each provides a way to prove this, but we are not there yet:

- ▶ The Grow-Algorithm Corollary implies that, if the Grow algorithm terminates, the set of vectors it has selected is a basis for the vector space \mathcal{V} .
However, we have not yet shown that it always terminates!
- ▶ The Shrink-Algorithm Corollary implies that, if we can run the Shrink algorithm starting with a finite set of vectors that spans \mathcal{V} , upon termination it will have selected a basis for \mathcal{V} .
However, we have not yet shown that every vector space \mathcal{V} is spanned by some finite set of vectors!

Computational problems involving finding a basis

Two natural ways to specify a vector space \mathcal{V} :

1. Specifying generators for \mathcal{V} .
2. Specifying a homogeneous linear system whose solution set is \mathcal{V} .

Two Fundamental Computational Problems:

Computational Problem: *Finding a basis of the vector space spanned by given vectors*

- ▶ *input:* a list $[\mathbf{v}_1, \dots, \mathbf{v}_n]$ of vectors
- ▶ *output:* a list of vectors that form a basis for $\text{Span} \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$.

Computational Problem: *Finding a basis of the solution set of a homogeneous linear system*

- ▶ *input:* a list $[\mathbf{a}_1, \dots, \mathbf{a}_n]$ of vectors
- ▶ *output:* a list of vectors that form a basis for the set of solutions to the system $\mathbf{a}_1 \cdot \mathbf{x} = 0, \dots, \mathbf{a}_n \cdot \mathbf{x} = 0$