# Linear Independence

Geometric Algorithms
Lecture 6

# Recap Problem

Do these three vectors span all of  $\mathbb{R}^3$ ?

$$\mathbf{v}_1 = \begin{bmatrix} -4 \\ 4 \\ 2 \end{bmatrix} \qquad \mathbf{v}_2 = \begin{bmatrix} -3 \\ 6 \\ -3 \end{bmatrix} \qquad \mathbf{v}_3 = \begin{bmatrix} -5 \\ 8 \\ -2 \end{bmatrix}$$

Consider the matrix

$$\begin{bmatrix} -4 & -3 & -5 \\ 4 & 6 & 8 \\ 2 & -3 & -2 \end{bmatrix}$$

$$\begin{bmatrix} -4 & -3 & -5 \\ 4 & 6 & 8 \\ 4 & -6 & -4 \end{bmatrix}$$

 $R_3 \leftarrow 2R_3$ 

$$\begin{bmatrix}
 -4 & -3 & -5 \\
 0 & 3 & 3 \\
 0 & -9 & -9
 \end{bmatrix}$$

$$R_2 \leftarrow R_2 + R_1$$

$$R_3 \leftarrow R_3 + R_1$$

$$\begin{bmatrix} -4 & -3 & -5 \\ 0 & 3 & 3 \\ 0 & 0 & 0 \end{bmatrix}$$

$$R_3 \leftarrow R_3 + 3R_2$$

$$\begin{bmatrix} -4 & -3 & -5 \\ 0 & 3 & 3 \\ 0 & 0 & 0 \end{bmatrix}$$

Third row has no pivot

# Objectives

- 1. Motivation
- 2. Define linear independence
- 3. See several perspectives on linear independence

# Keywords

linear independence
linear dependence
homogenous systems of linear equations
trivial and nontrivial solutions

# Motivation

#### Recall: Number of Solutions

zero the system is inconsistent

one the system has a unique solution

many the system has infinity solutions

# Up to Now

```
system of linear equations consistent?
     vector equation
    matrix equation
```

Does it have more than zero solutions?

Check the echelon form for consistency

Check the reduced echelon form for a solution

#### Some New Questions

When does Ax = b have exactly one solution?

When does Ax = b have infinitely many solutions?

What does it mean geometrically in each case?

# Homogeneous Linear Systems

#### Recall: The Zero Vector

# Homogenous Linear Systems

**Definition.** A system of linear equations is called *homogeneous* if it can be expressed as

$$A\mathbf{x} = \mathbf{0}$$



# Homogenous Linear Systems

**Definition.** A system of linear equations is called *homogeneous* if it can be expressed as

$$x_1 \mathbf{a}_1 + x_2 \mathbf{a}_2 + \dots + x_n \mathbf{a}_n = \mathbf{0}$$



# Homogenous Linear Systems

**Definition.** A system of linear equations is called *homogeneous* if it can be expressed as

$$a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = 0$$
 $a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = 0$ 
 $\vdots$ 
 $a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n = 0$ 
familiar...

(anxiety inducing?)

#### **Trivial Solutions**

Definition. For the matrix equation

$$A\mathbf{x} = \mathbf{0}$$

the solution x = 0 is called the **trivial solution**.

Any other solution is called *nontrivial*.

#### **Trivial Solutions**

Definition. For the vector equation

$$x_1 \mathbf{a}_1 + x_2 \mathbf{a}_2 + \dots + x_n \mathbf{a}_n = \mathbf{0}$$

the solution x = 0 is called the **trivial solution**.

Any other solution is called *nontrivial*.

#### **Trivial Solutions**

Definition. For the system of linear equations

$$a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = 0$$

$$a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = 0$$

$$\vdots$$

$$a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n = 0$$

the solution x=0 is called the **trivial solution**. Any other solution is called **nontrivial**.

# Questions about Homogeneous Systems

When does  $A\mathbf{x} = \mathbf{0}$  have only the trivial solution?

When does  $A\mathbf{x} = \mathbf{0}$  have nontrivial solutions?

What does it mean *geometrically* in each case?

# Linear Independence

# Linear Independence

**Definition.** A set of vectors  $\{v_1, v_2, ..., v_n\}$  is **linearly independent** if the vectors equation

$$x_1\mathbf{v}_1 + x_2\mathbf{v}_2 + \dots + x_n\mathbf{v}_n = \mathbf{0}$$

has exactly one solution (the trivial solution).

The columns of A are linearly independent if  $A\mathbf{x} = \mathbf{0}$  has exactly one solution.

# Linear Dependence

**Definition.** A set of vectors  $\{\mathbf v_1, \mathbf v_2, ..., \mathbf v_n\}$  is **linearly dependent** if the vectors equation

$$x_1\mathbf{v}_1 + x_2\mathbf{v}_2 + \ldots + x_n\mathbf{v}_n = \mathbf{0}$$

has a nontrivial solution.

A set of vectors is linearly dependent if there is a nontrivial linear combination of the vectors which equals  $\mathbf{0}$ .

# Linear Dependence (Alternative)

**Definition.** A set of vectors is **linearly dependent** if it is <u>not</u> linearly independent.

 $A\mathbf{x} = \mathbf{0}$  has a nontrivial solution

 $A\mathbf{x} = \mathbf{0}$  does <u>not</u> have only the trivial solution

# Another Interpretation of Linear Dependence

# demo (from ILA)

### Three Vectors in $\mathbb{R}^3$

It's possible for three vectors in  $\mathbb{R}^3$  to span all of  $\mathbb{R}^3$ , but it's <u>not</u> guaranteed

There may be vectors which lies in the plane spanned by two other vectors.

Or even two vectors which lie in the span of one of the others.

#### Fundamental Concern

How do we classify when a set of vectors does <u>not</u> span as much as it possibly can? When it is "smaller" than it could be?

This is the role of linear dependence.

# Linear Dependence (Another Alternative)

**Definition.** A set of vectors  $\{\mathbf{v}_1, \mathbf{v}_2, ..., \mathbf{v}_n\}$  is **linearly dependent** if it is nonempty and one of its vectors can be written as a linear combination of the others (not including itself).

$$\mathbf{v}_1 = \begin{bmatrix} -4\\4\\2 \end{bmatrix}$$
  $\mathbf{v}_2 = \begin{bmatrix} -3\\6\\-3 \end{bmatrix}$   $\mathbf{v}_3 = \begin{bmatrix} -5\\8\\-2 \end{bmatrix}$ 

(the recap problem)

## The Linear Combination Perspective

Suppose we have four vectors such that

$$\mathbf{v}_3 = 2\mathbf{v}_1 + 3\mathbf{v}_2 + 5\mathbf{v}_4$$

This gives us a solution to a vector equation.

# The Linear Combination Perspective

$$\mathbf{v}_3 = 2\mathbf{v}_1 + 3\mathbf{v}_2 + 5\mathbf{v}_4$$

implies

$$x_1 \mathbf{v}_1 + x_2 \mathbf{v}_2 + x_3 \mathbf{v}_3 + x_4 \mathbf{v}_4 = \mathbf{0}$$

has a nontrivial solution:

$$(2,3,-1,5)$$

# The Vector Equation Perspective

Suppose

$$x_1$$
**v**<sub>1</sub> +  $x_2$ **v**<sub>2</sub> +  $x_3$ **v**<sub>3</sub> +  $x_4$ **v**<sub>4</sub> = **0**

has a nontrivial solution:

$$(\alpha_1, \alpha_2, \alpha_3, \alpha_4)$$

where, say,  $\alpha_2 \neq 0$ 

We can turn this into a linear combination.

# The Vector Equation Perspective

$$\alpha_1 \mathbf{v}_1 + \alpha_2 \mathbf{v}_2 + \alpha_3 \mathbf{v}_3 + \alpha_4 \mathbf{v}_4 = \mathbf{0}$$

# The Vector Equation Perspective

$$\alpha_1 \mathbf{v}_1 + \alpha_3 \mathbf{v}_3 + \alpha_4 \mathbf{v}_4 = -\alpha_2 \mathbf{v}_2$$

## The Vector Equation Perspective

This division only works because  $\alpha_2 \neq 0$ .

$$\frac{-\alpha_1}{\alpha_2}\mathbf{v}_1 + \frac{-\alpha_3}{\alpha_2}\mathbf{v}_3 + \frac{-\alpha_4}{\alpha_2}\mathbf{v}_4 = \mathbf{v}_2$$

We get one vector as a linear combination of the others.

#### In All

**Theorem.** A set of vectors is linearly dependent if and only if it is nonempty and at least one of its vectors can be written as a linear combination of the others.

P if and only if Q means
P implies Q and Q implies P

## Linear Dependence Relation

**Definition.** If  $\mathbf{v}_1, \mathbf{v}_2, ..., \mathbf{v}_n$  are linearly dependent, then a *linear dependence relation* is an equation of the form

$$\alpha_1 \mathbf{v}_1 + \alpha_2 \mathbf{v}_2 + \dots + \alpha_n \mathbf{v}_n = \mathbf{0}$$

A linear dependence relation witnesses the linear dependence.

## How To: Linear Dependence Relation

**Question.** Write down a linear dependence relation for the vectors  $\mathbf{v}_1, \mathbf{v}_2, ... \mathbf{v}_n$ .

**Solution.** Find a nontrivial solution to the equation

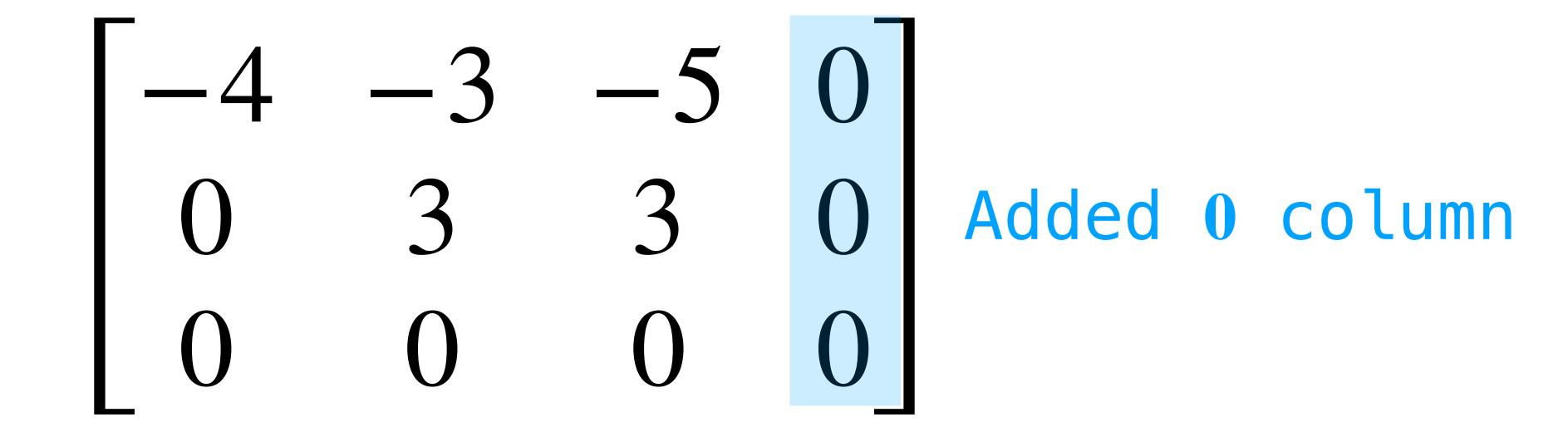
$$\begin{bmatrix} \mathbf{V}_1 & \mathbf{V}_2 & \dots & \mathbf{V}_n \end{bmatrix} \mathbf{x} = \mathbf{0}$$

(there will be a free variable you can choose to be nonzero)

## Question

Write down the linear dependence relation for the following vectors.

$$\mathbf{v}_1 = \begin{bmatrix} -4\\4\\2 \end{bmatrix} \qquad \mathbf{v}_2 = \begin{bmatrix} -3\\6\\-3 \end{bmatrix} \qquad \mathbf{v}_3 = \begin{bmatrix} -5\\8\\-2 \end{bmatrix}$$



Where we left off

$$\begin{bmatrix} -4 & -3 & -5 & 0 \\ 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

$$R_2 \leftarrow R_2/3$$

$$\begin{bmatrix} -4 & 0 & -2 & 0 \\ 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

$$R_1 \leftarrow R_1 + 3R_2$$

$$\begin{bmatrix} 1 & 0 & 0.5 & 0 \\ 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

$$R_1 \leftarrow R_1/(-4)$$

$$x_1 = -(0.5)x_3$$
 $x_2 = -x_3$ 
 $x_3$  is free

$$x_1 = 1$$
 $x_2 = 2$ 
 $x_3 = -2$ 

$$\begin{bmatrix} -4 \\ 4 \\ 2 \end{bmatrix} + 2 \begin{bmatrix} -3 \\ 6 \\ -3 \end{bmatrix} - 2 \begin{bmatrix} -5 \\ 8 \\ 2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

# Simple Cases

## The Empty Set

{} (a.k.a. Ø) is linearly independent

We stretch the definition a bit: there is no nontrivial linear combination of the vectors equaling  $\mathbf{0}$ 

There are none at all...

0 is in every span, even the empty span.

#### One Vector

A single vector  $\mathbf{v}$  is linearly independent if and only if it  $\mathbf{v} \neq \mathbf{0}.$ 

Note that

$$x_1 \mathbf{0} = \mathbf{0}$$

has many nontrivial solutions.

## The Zero Vector and Linear Dependence

If a set of vectors V contains the  $\mathbf{0}$ , then it is linearly dependent.

$$(1)\mathbf{0} + 0\mathbf{v}_2 + 0\mathbf{v}_2 + \dots + 0\mathbf{v}_n = \mathbf{0}$$

There is a very simple nontrivial solution.

#### **Two Vectors**

Definition. Two vectors are colinear if they are

scalar multiples of each other.

e.g., 
$$\begin{bmatrix} 1\\1\\2 \end{bmatrix}$$
 and  $\begin{bmatrix} 1.5\\1.5\\3 \end{bmatrix}$  or  $\begin{bmatrix} 2\\2 \end{bmatrix}$  and  $\begin{bmatrix} -1\\-1 \end{bmatrix}$ 

Two vectors are linearly dependent if and only if they are colinear.

#### Three Vectors

**Definition.** A collection of vectors is **coplanar** if their span is a plane.

Three vectors are linearly dependent if an only if they are colinear or coplanar.

This can be reasoning can be extended to more vectors, but we run out of terminology

## Yet Another Interpretation

## Increasing Span Criterion

If  $\mathbf{v}_1, \mathbf{v}_2, ..., \mathbf{v}_n$  are linearly *independent* then we cannot write one of it's vectors as a linear combination of the others.

But we get something stronger.

## Increasing Span Criterion

**Theorem.**  $\mathbf{v}_1, \mathbf{v}_2, ..., \mathbf{v}_n$  are linearly independent if and only if for all  $i \leq n$ ,

$$\mathbf{v}_i \not\in \operatorname{span}\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_{i-1}\}$$

As we add vectors, the span gets larger.

## Increasing Span Criterion

```
span{} is a point {0}
span\{v_1\} is a line
span\{v_1, v_2\} is a plane
span\{v_1, v_2, v_3\} is a 3-space
span\{v_1,v_2,v_3,v_4\} \ is \ a \ 4-space
```

## Characterization of Linear Dependence

**Theorem.**  $\mathbf{v}_1, \mathbf{v}_2, ..., \mathbf{v}_n$  are linearly dependent if and only there is an  $i \leq n$ ,

$$\mathbf{v}_i \in \text{span}\{\mathbf{v}_1, \mathbf{v}_2, ..., \mathbf{v}_{i-1}\}$$

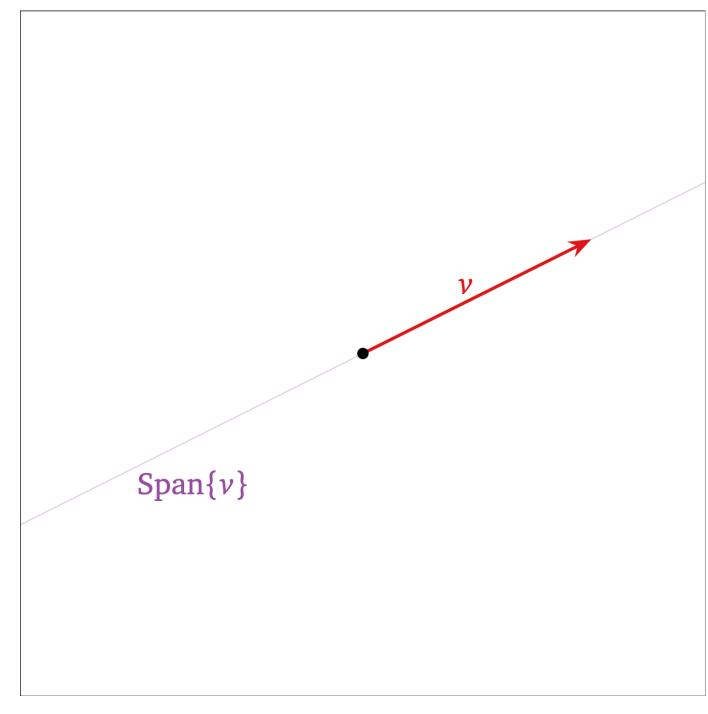
As we add vectors, we'll eventually find one in the span of the preceding ones.

## Characterization of Linear Dependence

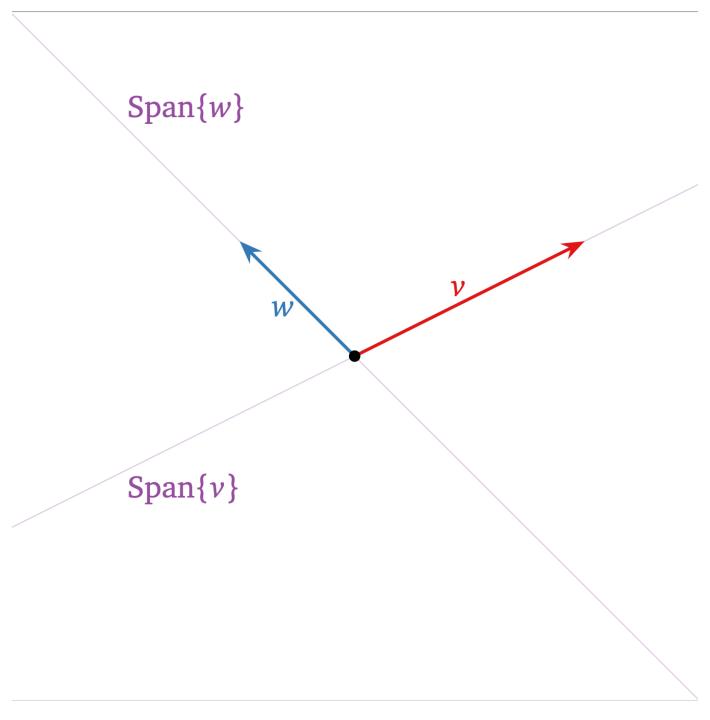
```
span{} is a point \{\mathbf{0}\} span\{\mathbf{v}_1\} is a line span\{\mathbf{v}_1,\mathbf{v}_2\} is a plane span\{\mathbf{v}_1,\mathbf{v}_2,\mathbf{v}_3\} is still a plane
```

 (this is an example, it may take a lot more vectors before we find one in the span of the preceding vectors)

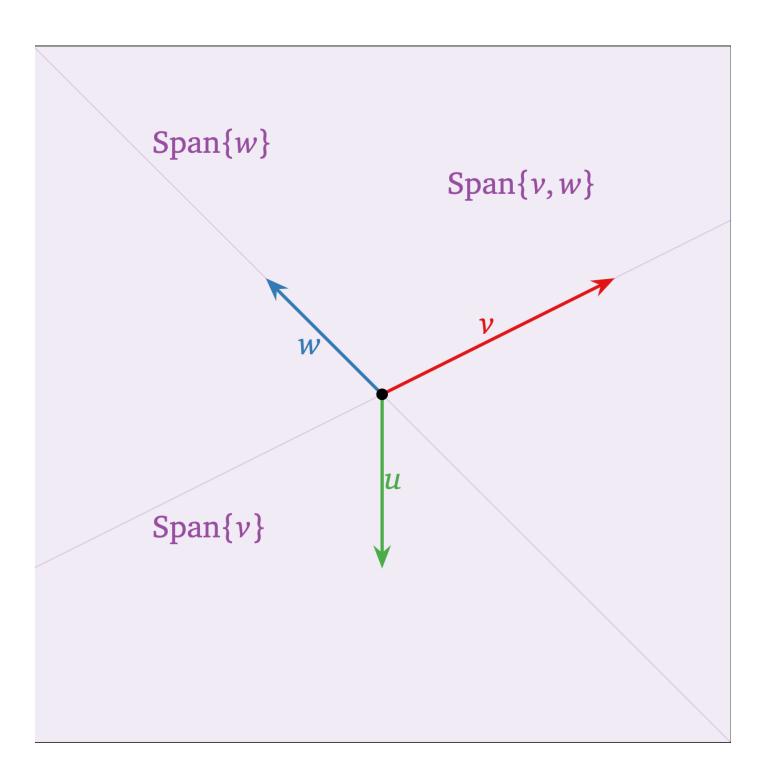
## As a Picture



span of 1 vector a line



span of 2 vector a plane



span of 3 vector still a plane

## Characterization of Linear Dependence

**Corollary.** If  $\mathbf{v}_1, \mathbf{v}_2, ..., \mathbf{v}_k$  are linearly dependent, then for any vector  $\mathbf{v}_{k+1}$ , the vectors  $\mathbf{v}_1, \mathbf{v}_2, ..., \mathbf{v}_k, \mathbf{v}_{k+1}$  are linearly dependent.

If we add a vector to a linearly dependent set, it remains linearly dependent

## Question

Are the following vectors linearly independent?

$$\mathbf{v}_1 = \begin{bmatrix} 1 \\ 2 \end{bmatrix} \qquad \mathbf{v}_2 = \begin{bmatrix} 2023 \\ 0 \end{bmatrix} \qquad \mathbf{v}_3 = \begin{bmatrix} 0.1 \\ 7 \end{bmatrix}$$

## Answer: No

$$\mathbf{v}_1 = \begin{bmatrix} 1 \\ 2 \end{bmatrix} \qquad \mathbf{v}_2 = \begin{bmatrix} 2023 \\ 0 \end{bmatrix} \qquad \mathbf{v}_3 = \begin{bmatrix} 0.1 \\ 7 \end{bmatrix}$$

Any three vectors can at most span a plane.

The first two are not colinear, so they span a plane  $(\mathbb{R}^2)$ .

# Linear Independence and Free Variables

## Linear Dependence Relations (Again)

When finding a linear dependence relation, we came across a system which a free variable

$$\begin{bmatrix} -4 & -3 & -5 & 0 \\ 0 & 3 & 3 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

we can take  $x_3$  to be free

## Pivots and Linear Dependence

**Theorem.** The columns of a matrix A are linearly independent if and only if A has a pivot in every <u>column</u>.

Remember that we choose our free variables to be the ones whose columns don't have pivots.

Free variables allow for infinitely many (nontrivial) solution.

$$x_1 = -(0.5)x_3$$
 $x_2 = -x_3$ 
 $x_3$  is free

$$x_1 = -0.5$$
 $x_2 = -1$ 
 $x_3 = 1$ 

$$x_1 = 0.5$$
 $x_2 = 1$ 
 $x_3 = -1$ 

$$x_1 = 1$$
 $x_2 = 2$ 
 $x_3 = -2$ 

The point: the solution is not unique.

## How To: Linear Independence

**Question.** Is the set of vectors  $\{a_1, a_2, ..., a_n\}$  linearly independent?

**Solution.** Check if  $x_1\mathbf{a}_1 + x_2\mathbf{a}_2 + \dots + x_n\mathbf{a}_n = \mathbf{0}$  has a unique solution.

### How To: Linear Independence

**Question.** Is the set of vectors  $\{a_1, a_2, ..., a_n\}$  linearly independent?

**Solution.** Check if  $[\mathbf{a}_1 \ \mathbf{a}_2 \ \dots \ \mathbf{a}_n] \mathbf{x} = \mathbf{0}$  has a unique solution.

### How To: Linear Independence

**Question.** Is the set of vectors  $\{a_1, a_2, ..., a_n\}$  linearly independent?

**Solution.** Check if the general form solution  $[\mathbf{a}_1 \ \mathbf{a}_2 \ \dots \ \mathbf{a}_n \ \mathbf{0}]$  has any free variables.

### How To: Linear Independence

**Question.** Is the set of vectors  $\{a_1, a_2, ..., a_n\}$  linearly independent?

**Solution.** Reduce  $[\mathbf{a}_1 \ \mathbf{a}_2 \ \dots \ \mathbf{a}_n]$  to echelon form and check if has a pivot position in every column.

### Example: Recap Problem Again

$$\mathbf{v}_1 = \begin{bmatrix} -4\\4\\2 \end{bmatrix} \qquad \mathbf{v}_2 = \begin{bmatrix} -3\\6\\-3 \end{bmatrix} \qquad \mathbf{v}_3 = \begin{bmatrix} -5\\8\\-2 \end{bmatrix}$$

The reduced echelon form of  $[v_1 \ v_2 \ v_3]$  is

$$\begin{bmatrix} 1 & 0 & 0.5 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \end{bmatrix}$$
 row without a pivot

### Linear Independence and Full Span

The columns of a matrix are linearly independent if there is a pivot in every <u>row</u>.

The columns of a matrix are linearly independent if there is a pivot in every <u>column</u>.

#### Tall Matrices

If m > n then the columns cannot span  $\mathbb{R}^m$ 

#### Tall Matrices

If m > n then the columns cannot span  $\mathbb{R}^m$ 

This matrix has at most 3 pivots, but 4 rows.

#### Wide Matrices

If m < n then the columns cannot be linearly independent

#### Wide Matrices

If m < n then the columns cannot be linearly independent

```
      1
      2
      3
      4

      5
      6
      7
      8

      9
      10
      11
      12
```

This matrix as at most 3 pivots, but 4 columns.

## The Takeaway

The columns of a matrix are linearly independent if there is a pivot in every <u>row</u>.

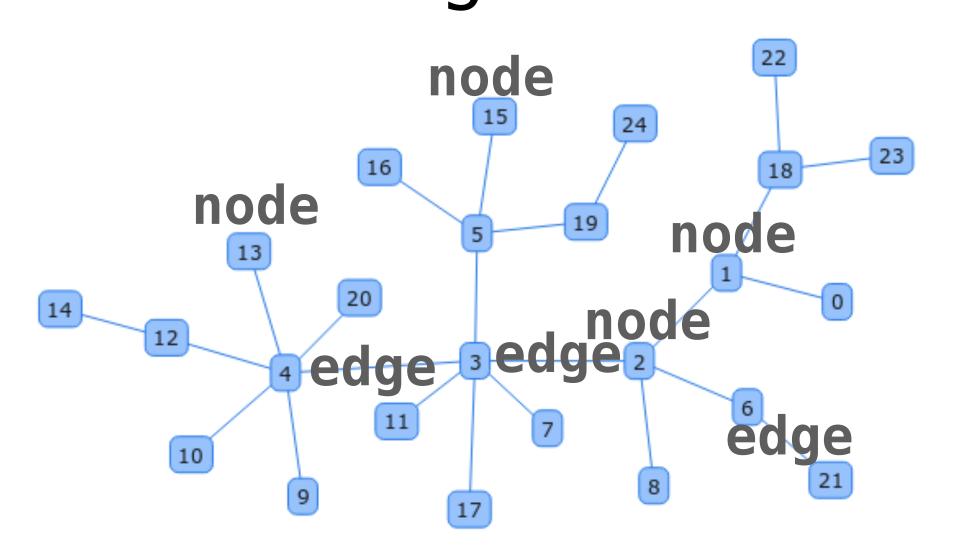
The columns of a matrix are linearly independent if there is a pivot in every <u>column</u>.

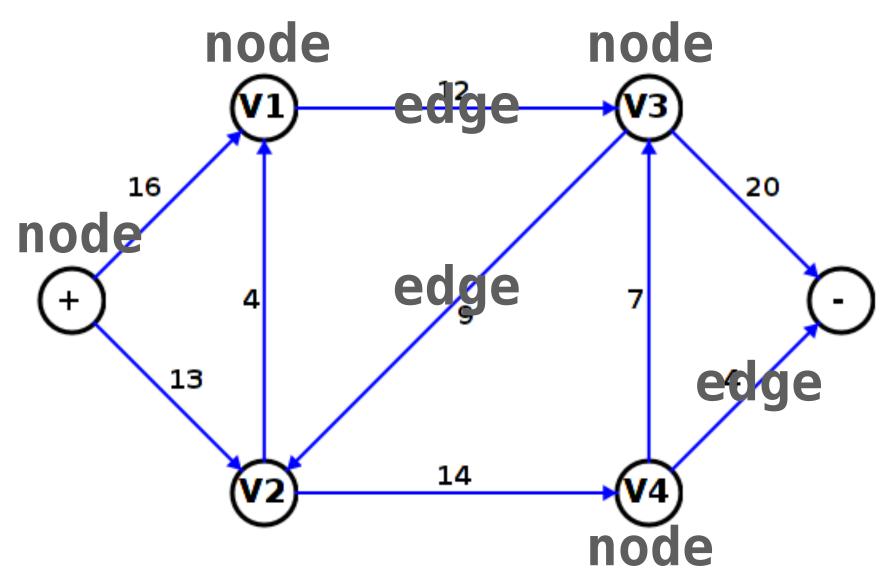
Don't confuse these!

# Application: Networks and Flow

### Graphs/Networks

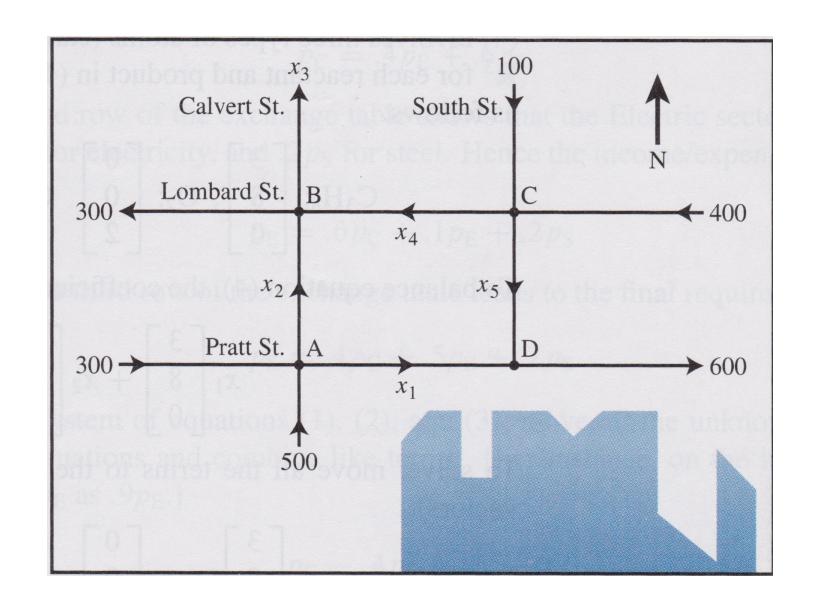
A *graph/network* is a mathematical object representing collection of *nodes* and *edges* connecting them.





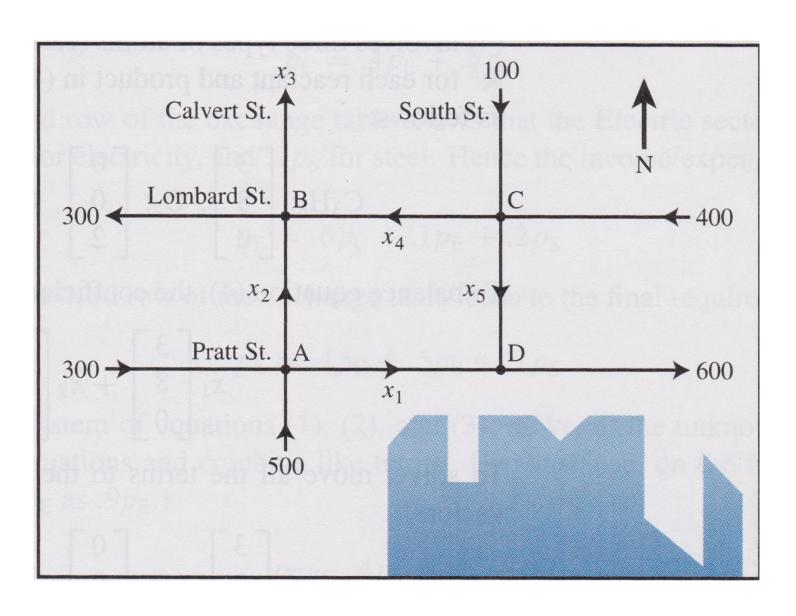
## Directed Graphs

Today we focus on *directed* graphs, in which edges have a specified direction.



Think of these as one-way streets.

#### Flow



We are often interested in how much "stuff" we can push through the edges

In the above example, the "stuff" is cars/hr.

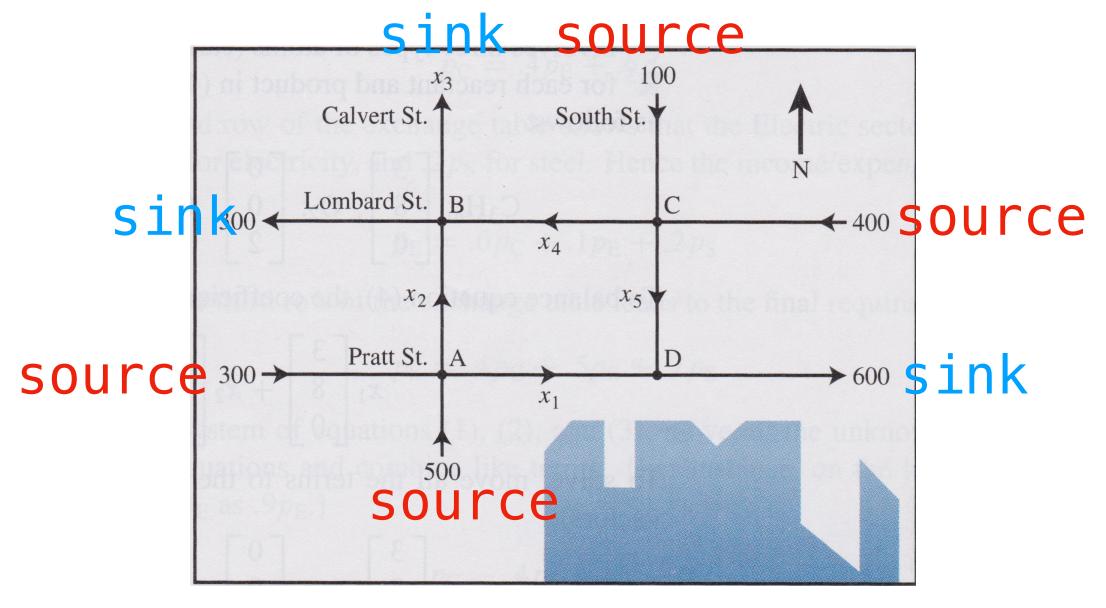
I like to imagine water moving through a pipe, and splitting an joints in the pipe

#### Flow Network

A flow network is a directed graph with specified source and sink nodes.

Flow <u>comes out of</u> and <u>goes into</u> sources and sinks. They are assigned a flow value (or

variable).



#### Flow

**Definition.** The **flow** of a graph is an assignment of <u>positive</u> values to the edges so that the following holds.

conservation: flow into a node = flow out of a node

source/sink constraint: flow into a source/out
of a sink is positive.

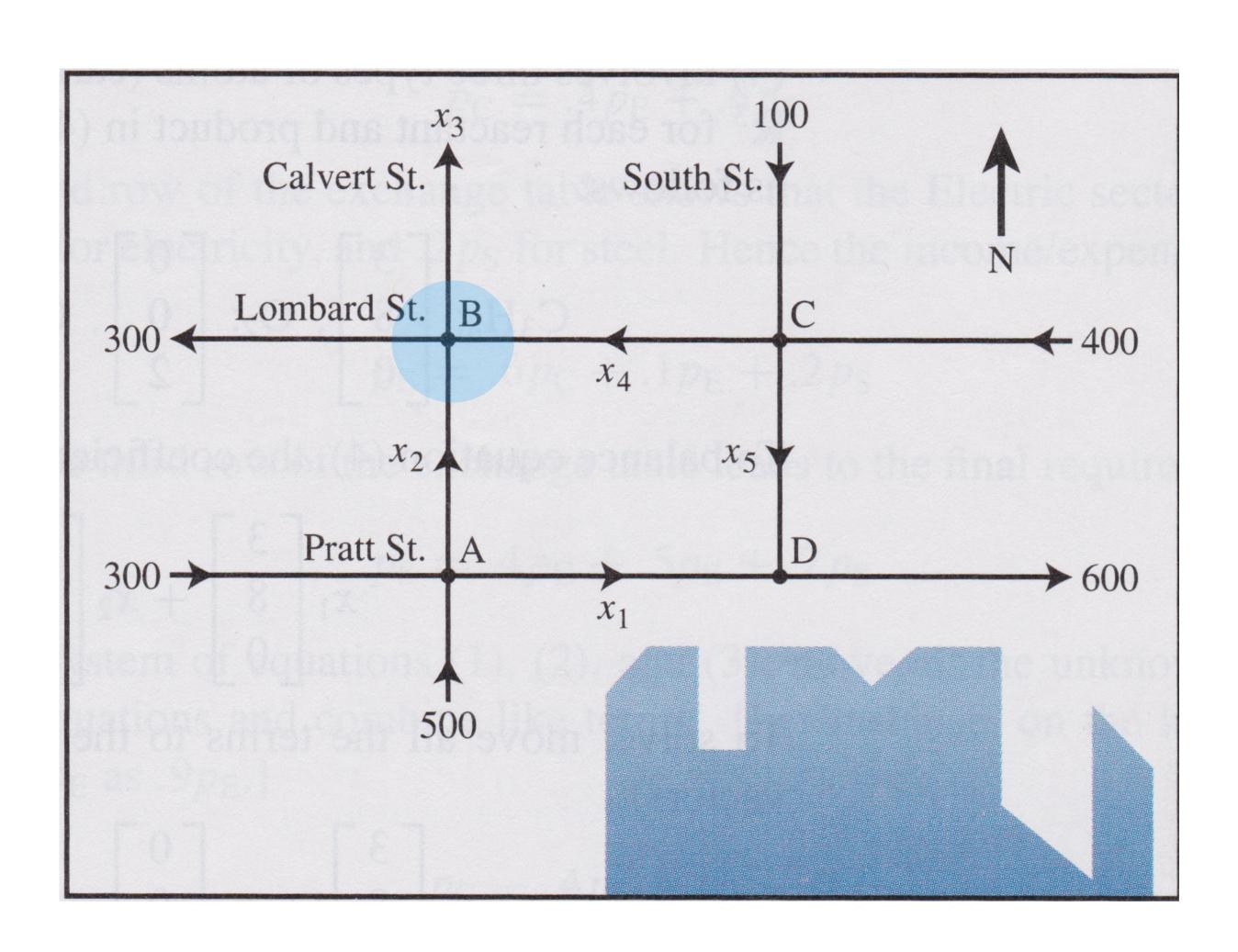
#### Flow Conservation

Flow in = Flow out

e.g.,

$$x_2 + x_4 = 300 + x_3$$

$$100 + 400 = x_4 + x_5$$



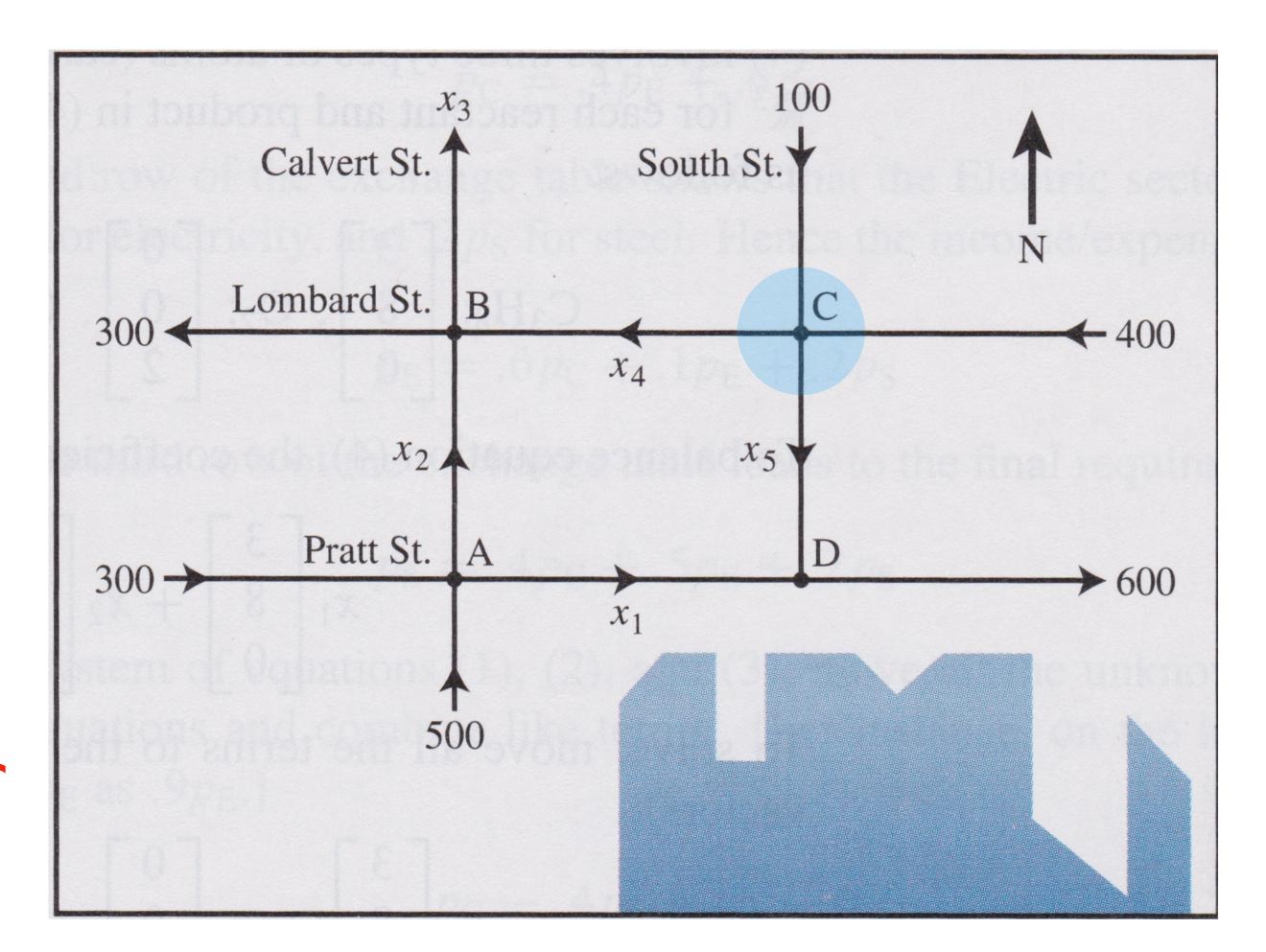
#### Flow Conservation

Flow in = Flow out

$$x_2 + x_4 = 300 + x_3$$

$$100 + 400 = x_4 + x_5$$

Every node determines a linear equation



#### How To: Network Flow

Question. Find a general solution for the flow of a given graph.

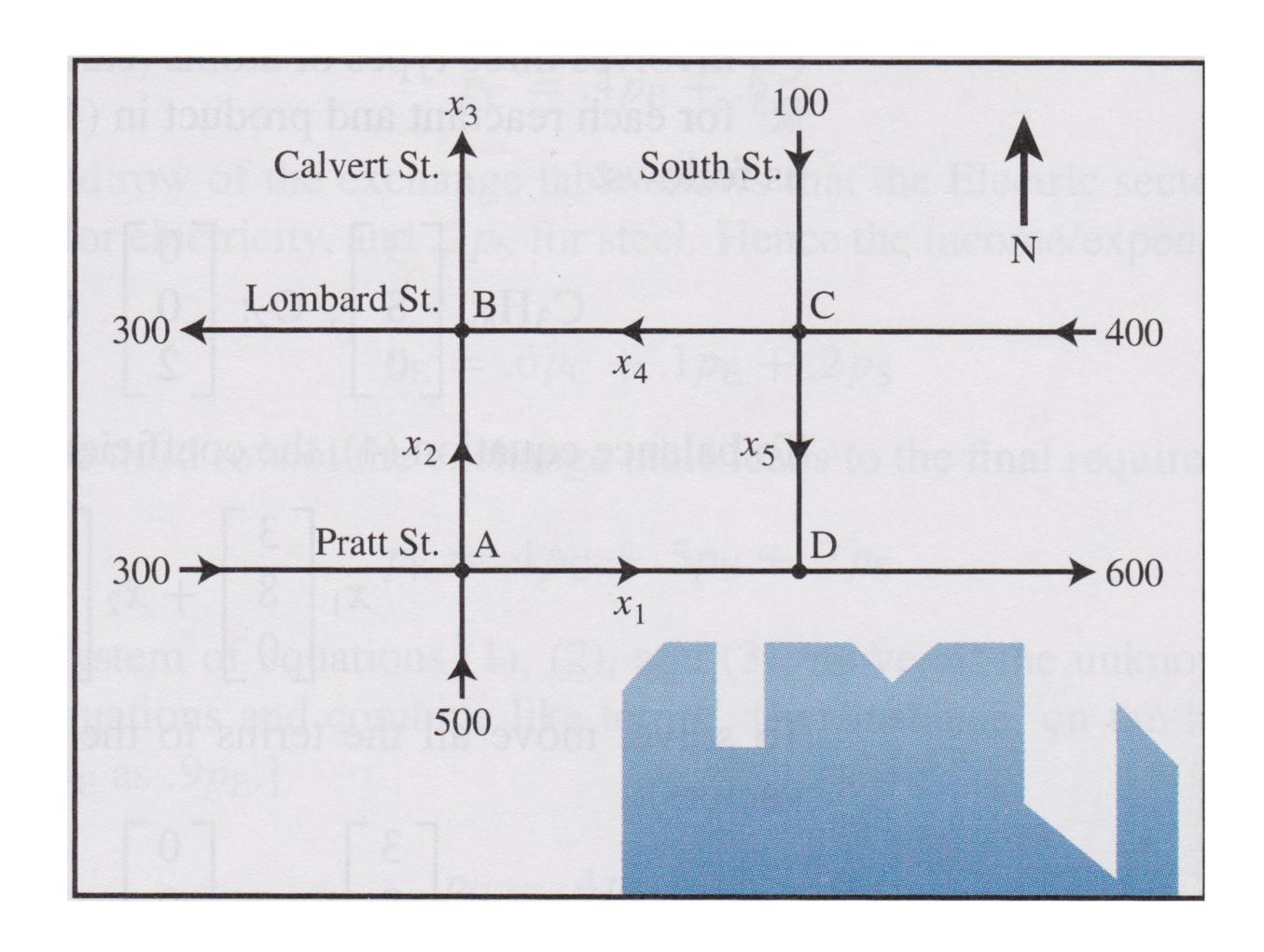
**Solution.** Write down the linear equations determined by <u>flow conservation</u> at non-source and non-sink nodes, and then solve.

(A) 
$$500 + 300 = x_1 + x_2$$

(B) 
$$x_2 + x_4 = 300 + x_3$$

(C) 
$$100 + 400 = x_4 + x_5$$

(D) 
$$x_1 + x_5 = 600$$



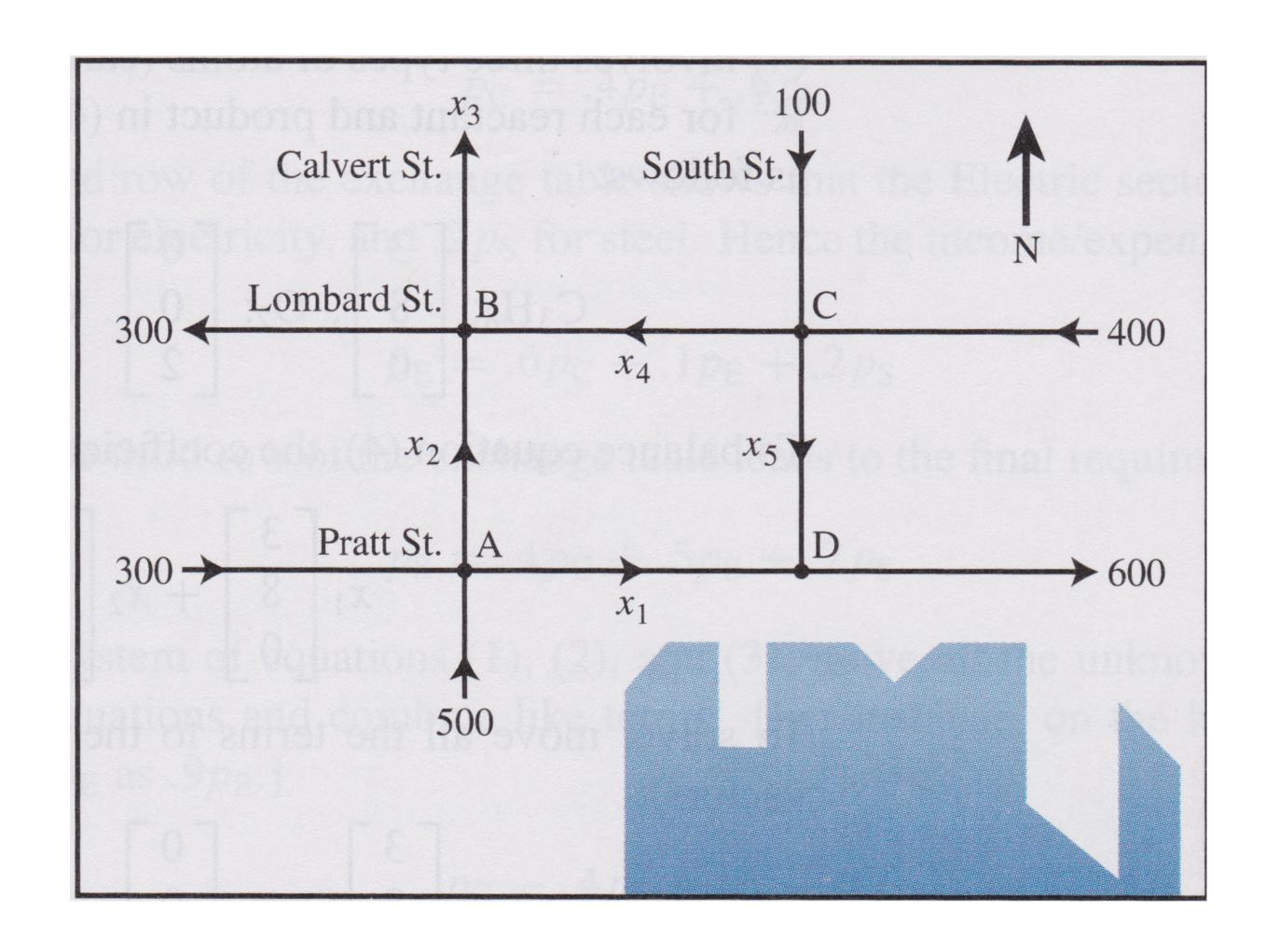
(A) 
$$x_1 + x_2 = 800$$

(B) 
$$x_2 - x_3 + x_4 = 300$$

(C) 
$$x_4 + x_5 = 500$$

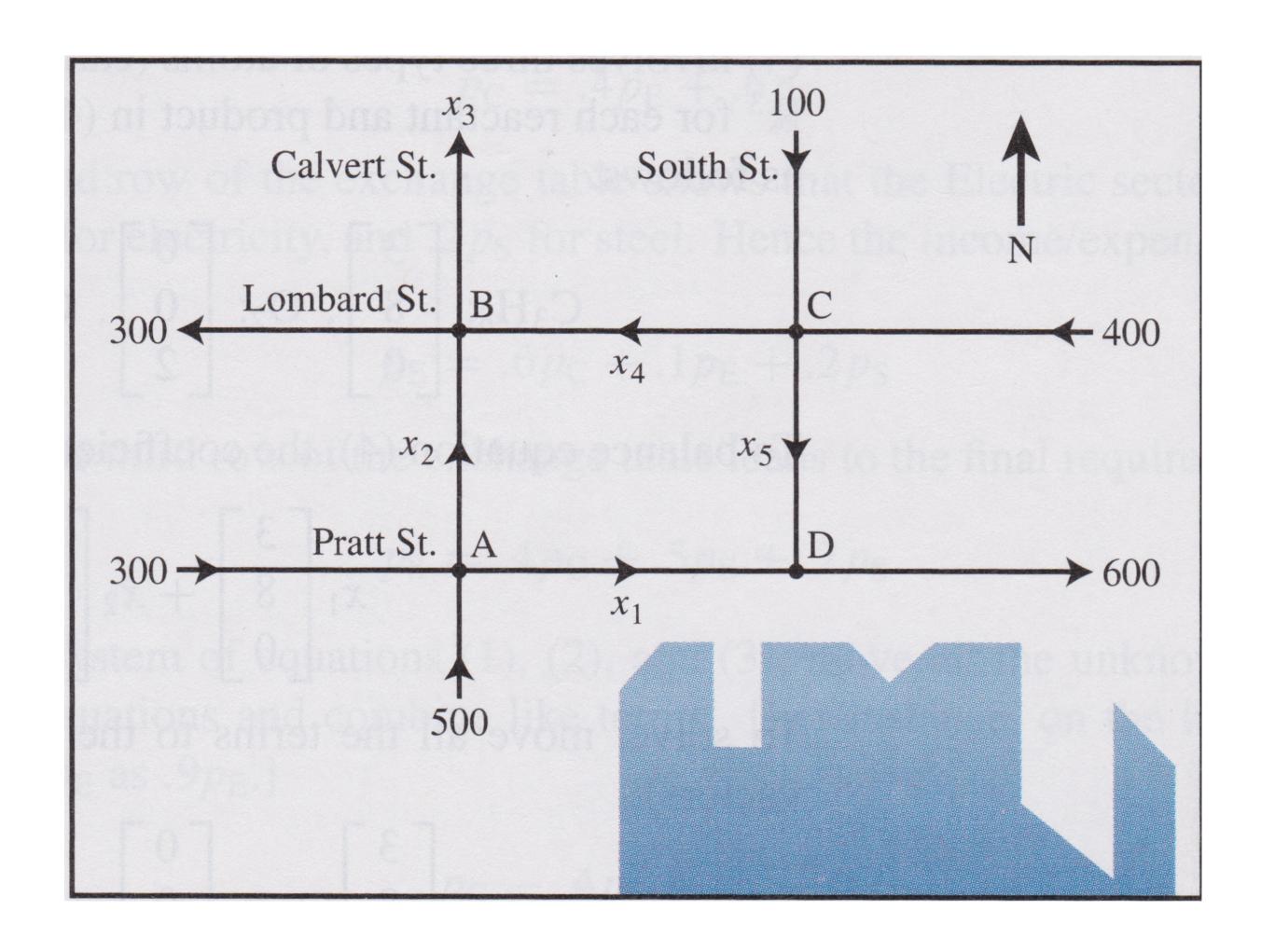
(D) 
$$x_1 + x_5 = 600$$

System of Linear Equations

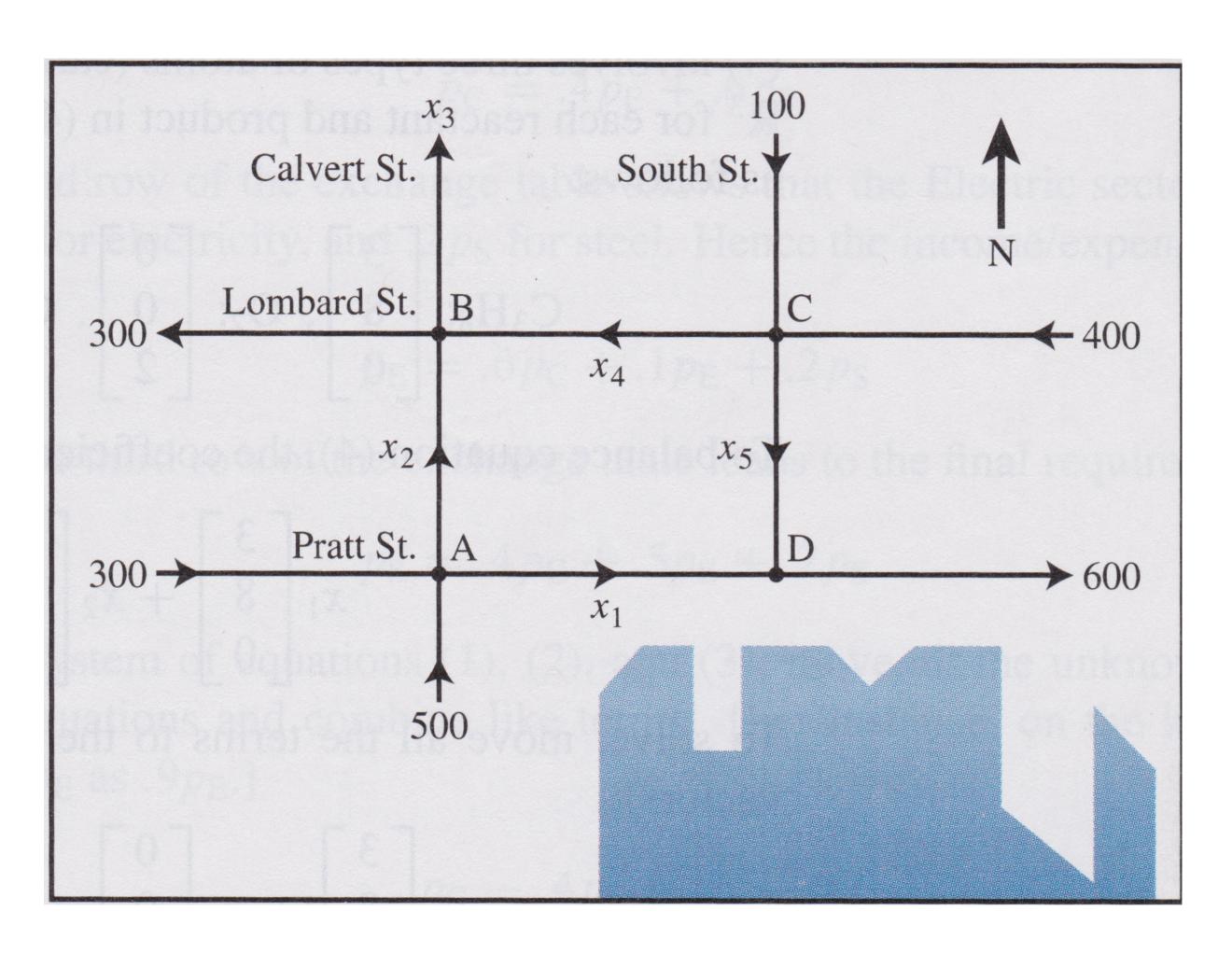


$$\begin{bmatrix} 1 & 1 & 0 & 0 & 0 & 800 \\ 0 & 1 & -1 & 1 & 0 & 300 \\ 0 & 0 & 0 & 1 & 1 & 500 \\ 1 & 0 & 0 & 0 & 1 & 600 \end{bmatrix}$$

Augmented Matrix



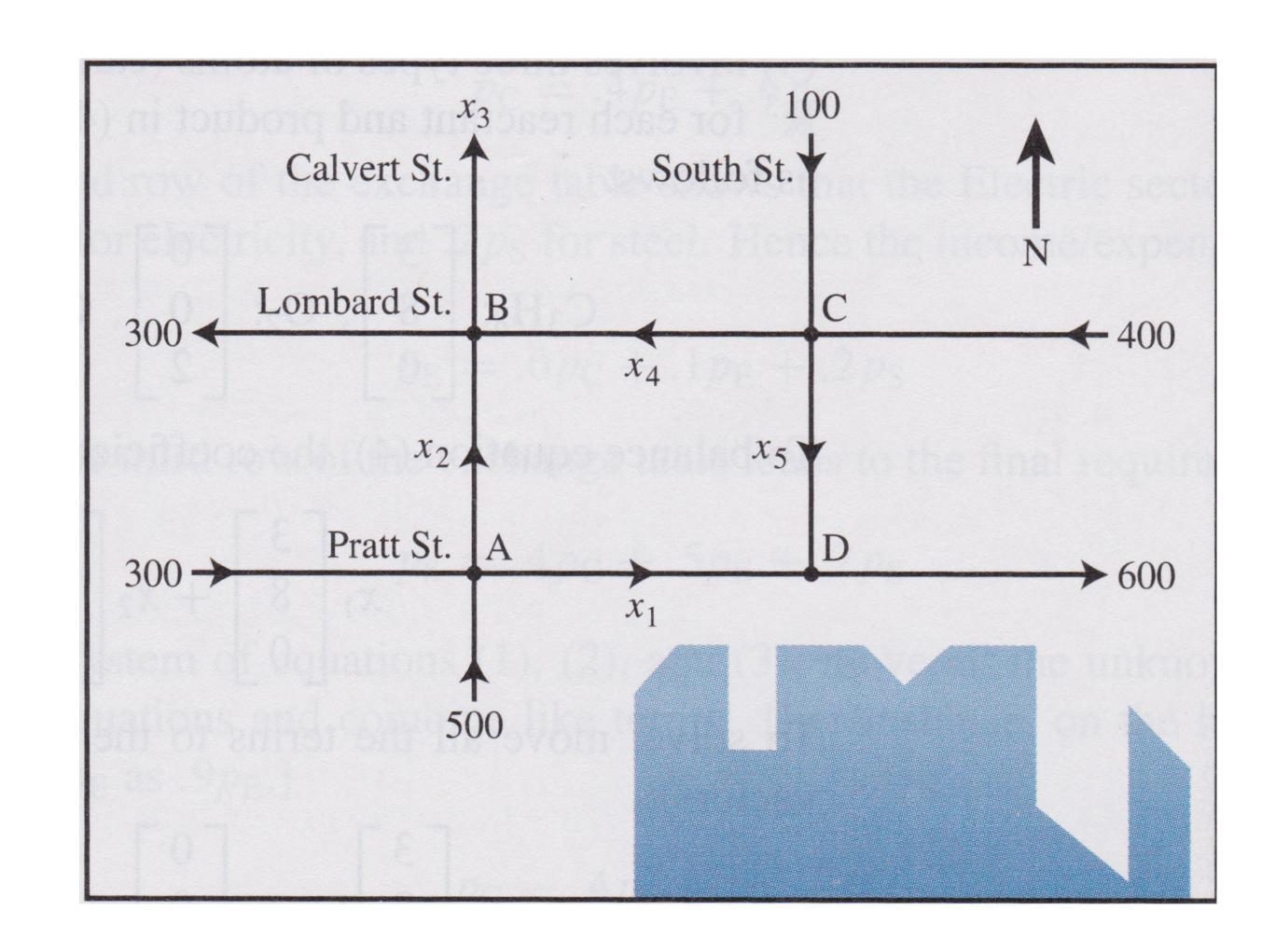
Reduced Echelon Form



Note that global flow is conserved.

$$x_1 = 600 - x_5$$
 $x_2 = 200 + x_5$ 
 $x_3 = 400$ 
 $x_4 = 500 - x_5$ 
 $x_5$  is free

General Solution



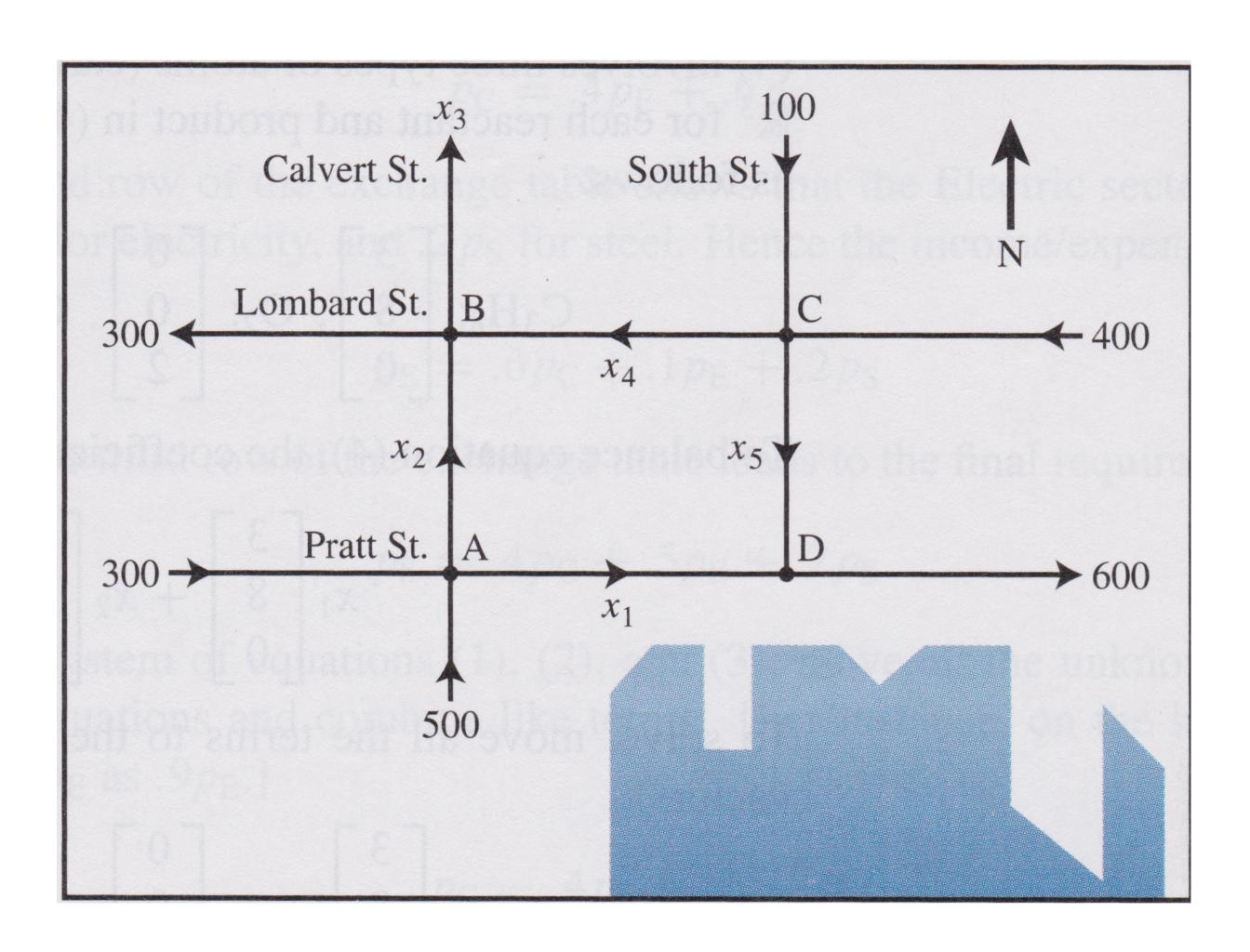
### How To: Max Flow Value for an Edge

Question. Find the maximum value of a flow variable in a given flow network.

**Solution.** Remember that flow values must be positive. Look at the general form solution and see what makes this hold.

$$x_1 = 600 - x_5$$
 $x_2 = 200 + x_5$ 
 $x_3 = 400$ 
 $x_4 = 500 - x_5$ 
 $x_5$  is free

$$x_4 \ge 0$$
 implies  $x_5 \le 500$   
 $x_1 \ge 0$  implies  $x_5 \le 600$ 



The maximum value of  $x_5$  is 500

### Summary

Linear independence helps us understand when a span is "as large as it can be."

We can reduce this seeing if a single homogeneous equation has a unique solution.

Network Flows define linear systems we can solve.