

CS 3100

Data Structures and Algorithms 2

Lecture 19: Longest Common Subsequence

Co-instructors: Robbie Hott and Ray Pettit
Spring 2024

Readings in CLRS 4th edition:

- Chapter 14

Announcements

- PS8 due Wednesday
- Quizzes 3-4 coming next week
- Office hours updates
 - Prof Hott Office Hours:
 - Today 4/2: 2-3pm
 - Back to normal starting Friday

Dynamic Programming

- Requires **Optimal Substructure**
 - Solution to larger problem contains the (optimal) solutions to smaller ones
- Idea:
 1. Identify the recursive structure of the problem
 - What is the “last thing” done?
 2. Save the solution to each subproblem in memory
 3. Select a good order for solving subproblems
 - “Top Down”: Solve each recursively
 - “Bottom Up”: Iteratively solve smallest to largest

Log Cutting

Given a log of length n

A list (of length n) of prices P ($P[i]$ is the price of a cut of size i)

Find the best way to cut the log

Price:	1	5	8	9	10	17	17	20	24	30
Length:	1	2	3	4	5	6	7	8	9	10



Select a list of lengths ℓ_1, \dots, ℓ_k such that:

$$\sum \ell_i = n$$

to maximize $\sum P[\ell_i]$

Brute Force: $O(2^n)$

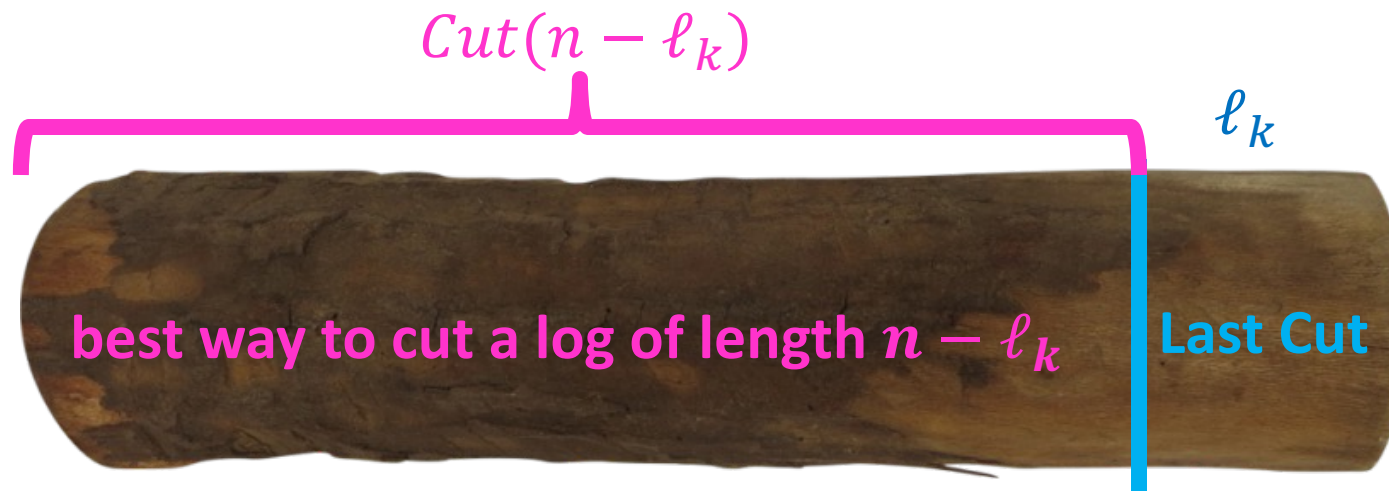
1. Identify Recursive Structure

$P[i]$ = value of a cut of length i

$Cut(n)$ = value of best way to cut a log of length n

$$Cut(n) = \max \begin{cases} Cut(n-1) + P[1] \\ Cut(n-2) + P[2] \\ \dots \\ Cut(0) + P[n] \end{cases}$$

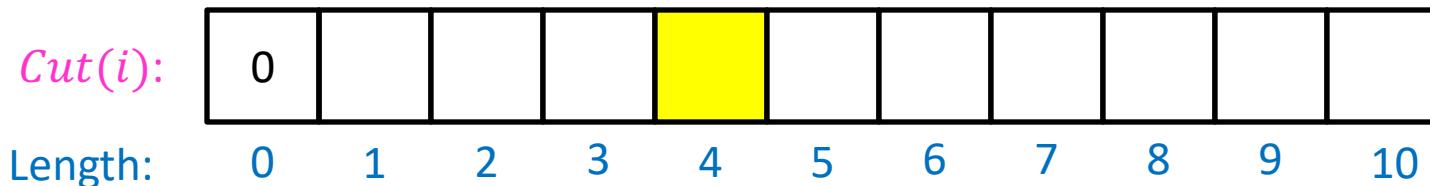
2. Save sub-solutions to memory!



3. Select a Good Order for Solving Subproblems

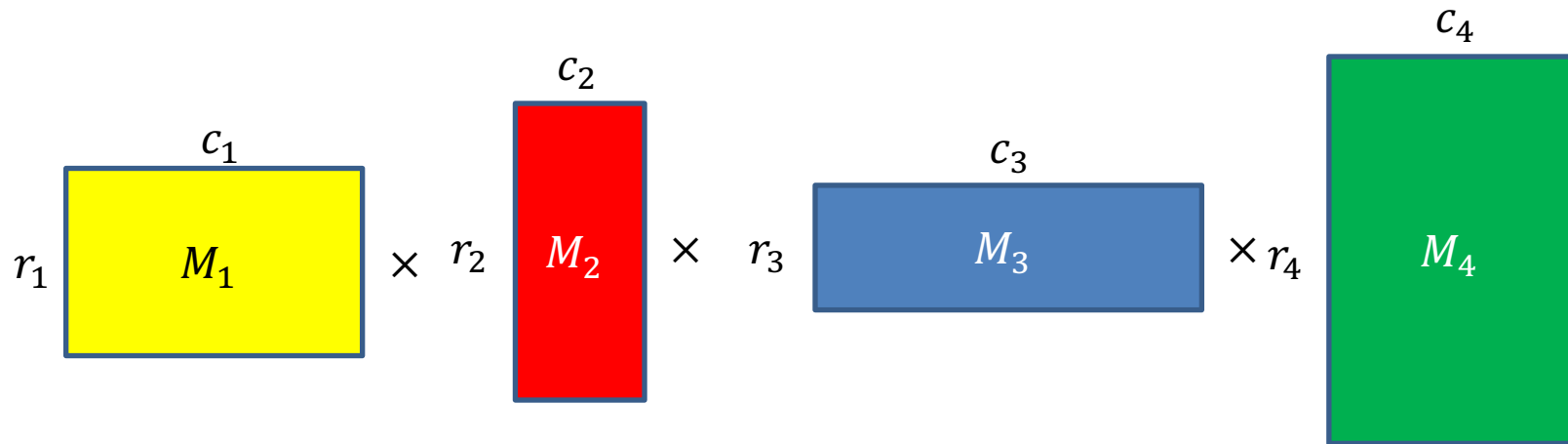
Solve Smallest subproblem first

$$Cut(4) = \max \begin{cases} Cut(3) + P[1] \\ Cut(2) + P[2] \\ Cut(1) + P[3] \\ Cut(0) + P[4] \end{cases}$$



Matrix Chaining

- Given a sequence of Matrices (M_1, \dots, M_n) , what is the most efficient way to multiply them?



1. Identify the Recursive Structure of the Problem

- In general:

$Best(i, j)$ = cheapest way to multiply together M_i through M_j

$$Best(i, j) = \min_{k=i}^{j-1} (Best(i, k) + Best(k + 1, j) + r_i r_{k+1} c_j)$$

$$Best(i, i) = 0$$

$$Best(1, n) = \min \left\{ \begin{array}{l} Best(2, n) + r_1 r_2 c_n \\ Best(1, 2) + Best(3, n) + r_1 r_3 c_n \\ Best(1, 3) + Best(4, n) + r_1 r_4 c_n \\ Best(1, 4) + Best(5, n) + r_1 r_5 c_n \\ \dots \\ Best(1, n - 1) + r_1 r_n c_n \end{array} \right.$$

2. Save Subsolutions in Memory

- In general:

$Best(i, j)$ = cheapest way to multiply together M_i through M_j

$$Best(i, j) = \min_{k=i}^{j-1} (Best(i, k) + Best(k+1, j) + r_i r_{k+1} c_j)$$

$$Best(i, i) = 0$$

Save to $M[n]$

Read from $M[n]$
if present

$$Best(1, n) = \min$$

$$Best(2, n) + r_1 r_2 c_n$$

$$Best(1, 2) + Best(3, n) + r_1 r_3 c_n$$

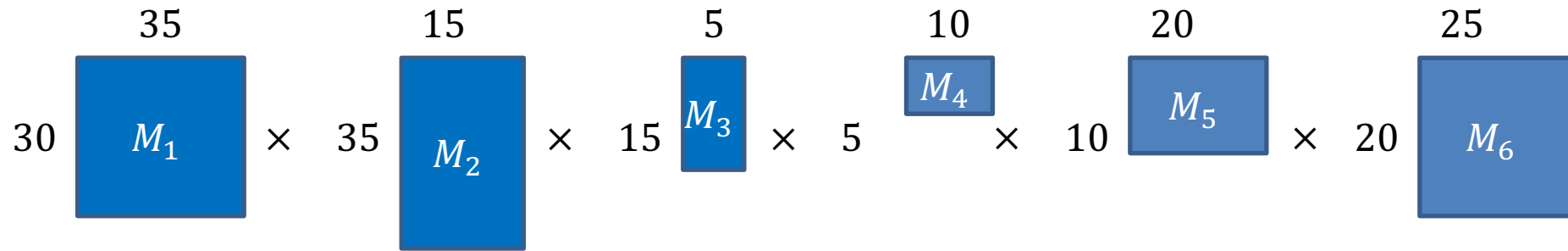
$$Best(1, 3) + Best(4, n) + r_1 r_4 c_n$$

$$Best(1, 4) + Best(5, n) + r_1 r_5 c_n$$

...

$$Best(1, n-1) + r_1 r_n c_n$$

3. Select a good order for solving subproblems



$$Best(i, j) = \min_{k=i}^{j-1} (Best(i, k) + Best(k + 1, j) + r_i r_{k+1} c_j)$$

$$Best(i, i) = 0$$

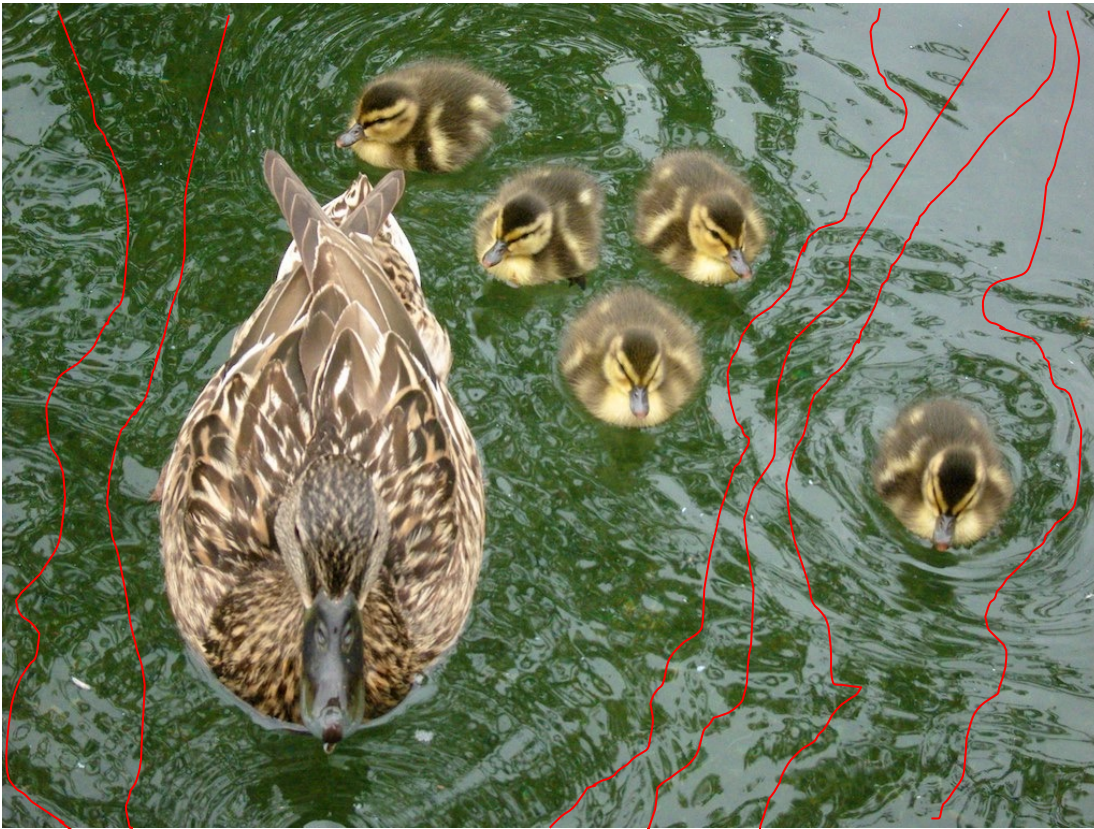
$j =$	1	2	3	4	5	6	$= i$
1	0	15750	7875				1
2		0	2625				2
3			0	750			3
4				0	1000		4
5					0	5000	5
6						0	6

To find $Best(i, j)$: Need all preceding terms of row i and column j

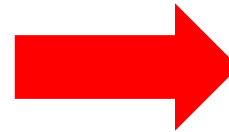
Conclusion: solve in order of diagonal

Seam Carving

- Removes “least energy seam” of pixels
- <https://trekhleb.dev/js-image-carver/>

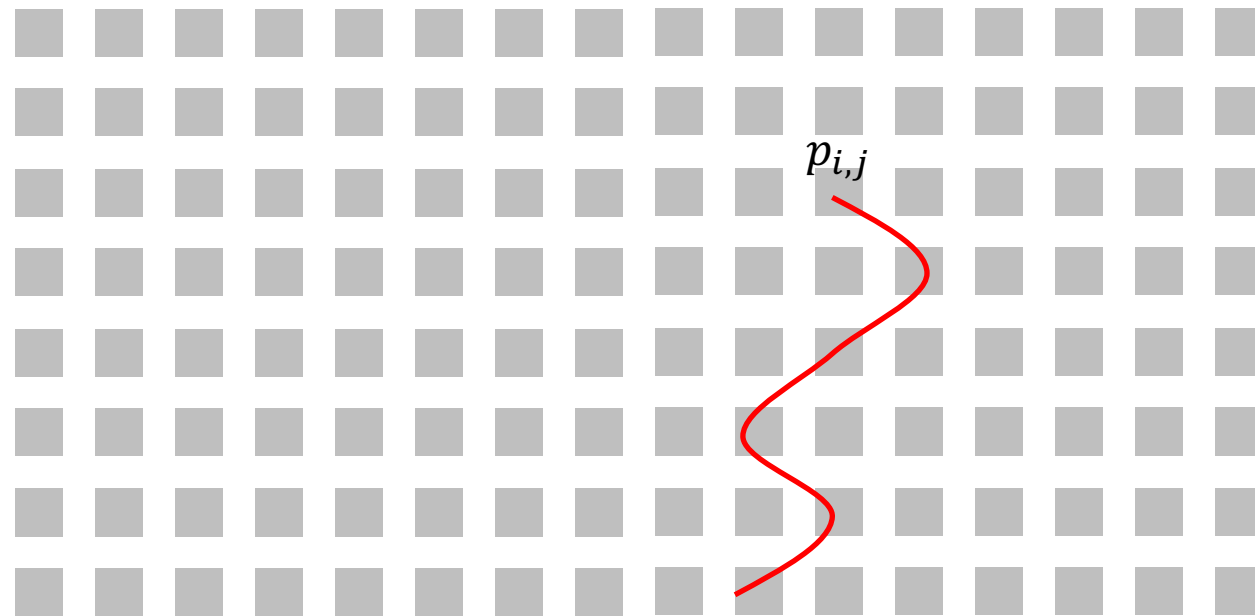


Carved



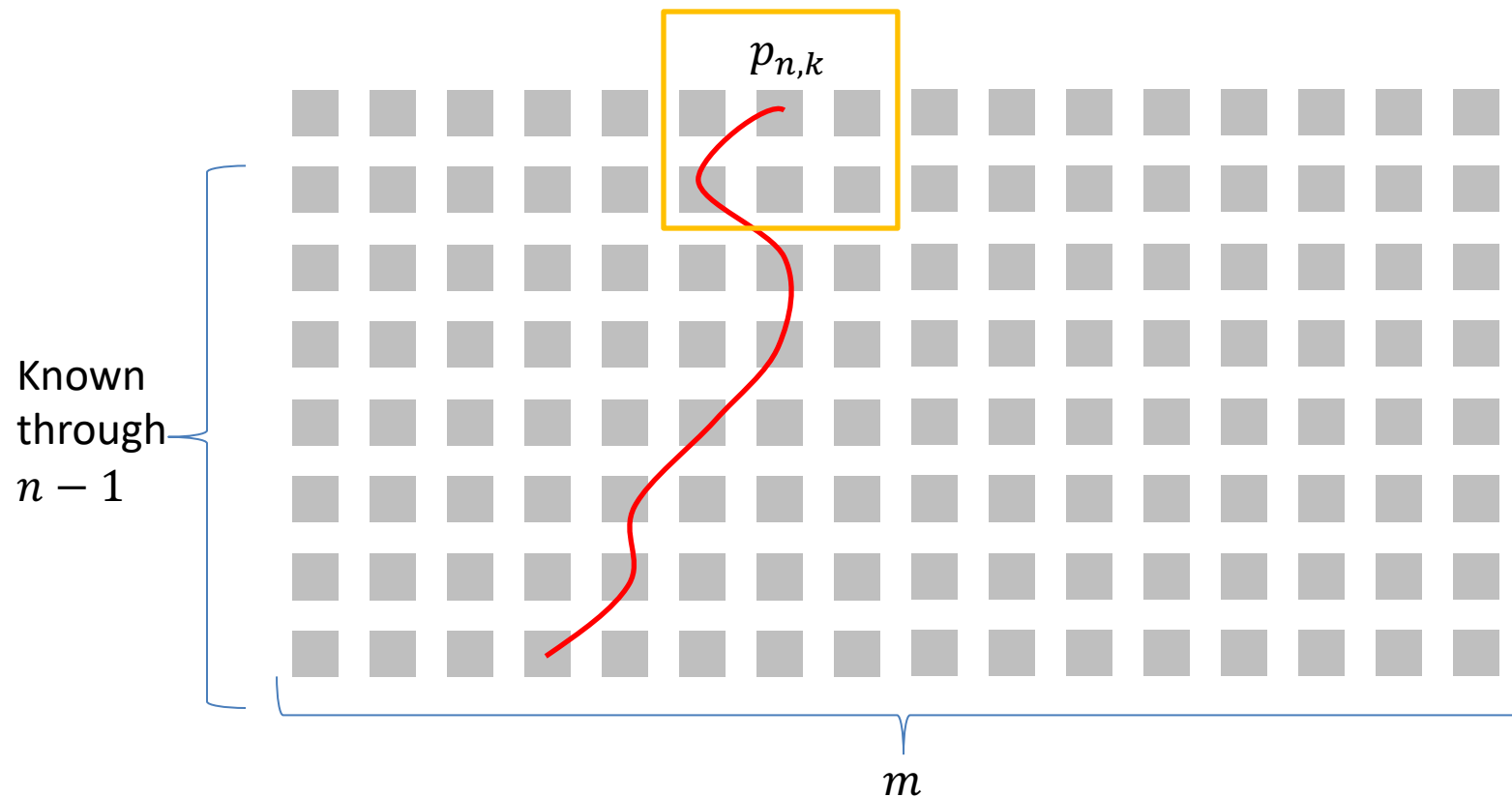
Identify Recursive Structure

Let $S(i, j)$ = least energy seam from the bottom of the image up to pixel $p_{i,j}$



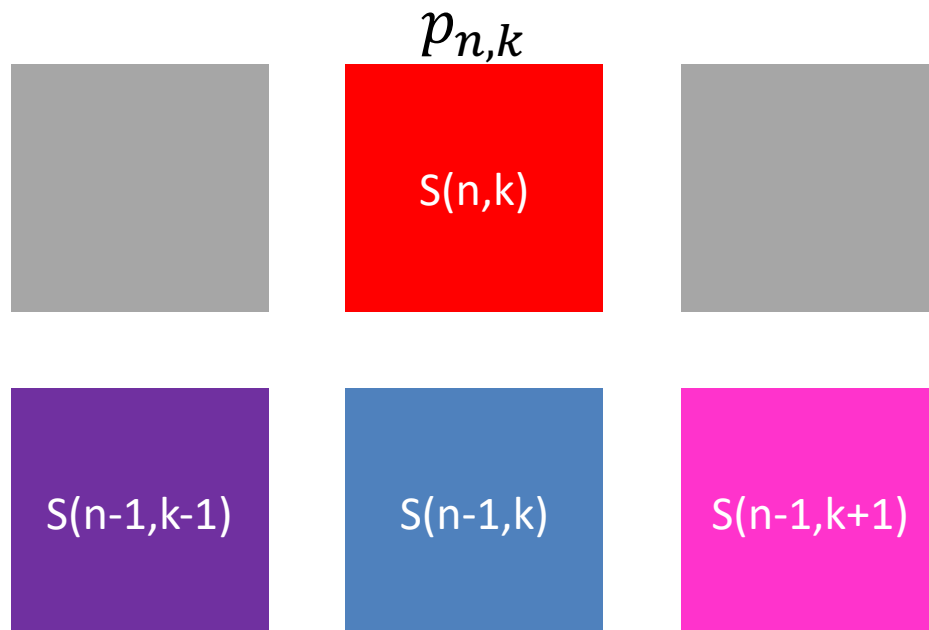
Computing $S(n, k)$

Assume we know the least energy seams for all of row $n - 1$
(i.e. we know $S(n - 1, \ell)$ for all ℓ)



Computing $S(n, k)$

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Computing $S(n, k)$

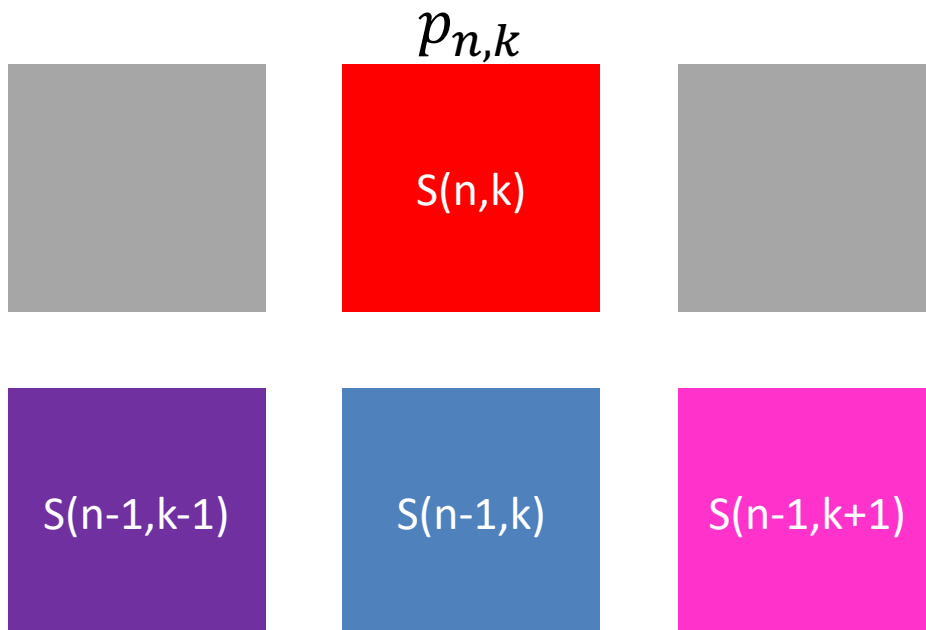
Assume we know the least energy seams for all of row $n - 1$ (i.e. we know $S(n - 1, \ell)$ for all ℓ)

$$S(n, k) = \min$$

$$S(n - 1, k - 1) + e(p_{n,k})$$

$$S(n - 1, k) + e(p_{n,k})$$

$$S(n - 1, k + 1) + e(p_{n,k})$$



Coin Changing: Identify Recursive Structure

Change(n): minimum number of coins needed to give change for n cents

Possibilities for last coin



Coins needed

$$\text{Change}(n - 25) + 1 \quad \text{if } n \geq 25$$

$$\text{Change}(n - 11) + 1 \quad \text{if } n \geq 11$$

$$\text{Change}(n - 10) + 1 \quad \text{if } n \geq 10$$

$$\text{Change}(n - 5) + 1 \quad \text{if } n \geq 5$$

$$\text{Change}(n - 1) + 1 \quad \text{if } n \geq 1$$

Identify Recursive Structure

Change(n): minimum number of coins needed to give change for n cents

$$\text{Change}(n) = \min \begin{cases} \text{Change}(n - 25) + 1 & \text{if } n \geq 25 \\ \text{Change}(n - 11) + 1 & \text{if } n \geq 11 \\ \text{Change}(n - 10) + 1 & \text{if } n \geq 10 \\ \text{Change}(n - 5) + 1 & \text{if } n \geq 5 \\ \text{Change}(n - 1) + 1 & \text{if } n \geq 1 \end{cases}$$

Correctness: The optimal solution must be contained in one of these configurations

Base Case: Change(0) = 0

Running time: $O(kn)$

k is number of possible coins

Is this efficient?

No, this is pseudo-polynomial time

Input size is $O(k \log n)$

Longest Common Subsequence

Given two sequences X and Y ,
find the length of their longest
common subsequence

Example:

$X = TGCATA$

$Y = ATCTGAT$

$LCS = TCTA$

Brute force: Compare every
subsequence of X with Y
 $\Omega(2^n)$



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1. Identify Recursive Structure

Let $LCS(i, j)$ = length of the LCS for the first i characters of X , first j character of Y

Find $LCS(i, j)$:

Case 1: $X[i] = Y[j]$

$X = \text{TGCATAT}$

$Y = \text{ATCTGCGT}$

$$LCS(i, j) = LCS(i - 1, j - 1) + 1$$

Case 2: $X[i] \neq Y[j]$

$X = \text{TGCATAC}$

$Y = \text{ATCTGCGT}$

$$LCS(i, j) = LCS(i, j - 1)$$

$X = \text{TGCATAT}$

$Y = \text{ATCTGCGA}$

$$LCS(i, j) = LCS(i - 1, j)$$

$$LCS(i, j) = \begin{cases} 0 & \text{if } i = 0 \text{ or } j = 0 \\ LCS(i - 1, j - 1) + 1 & \text{if } X[i] = Y[j] \\ \max(LCS(i, j - 1), LCS(i - 1, j)) & \text{otherwise} \end{cases}$$

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↑ Save to $M[i, j]$
↖ Read from $M[i, j]$ if present

Top-Down Solution with Memoization

We need two functions; one will be recursive.

LCS-Length(X, Y) // Y is M's cols.

1. $n = \text{length}(X)$
2. $m = \text{length}(Y)$
3. Create table $M[n,m]$
4. Assign -1 to all cells $M[i,j]$
- // get value for entire sequences
5. return **LCS-recur**(X, Y, M, n, m)

LCS-recur(X, Y, M, i, j)

1. if $(i == 0 \ || \ j == 0)$ return 0
- // have we already calculated this subproblem?
2. if $(M[i,j] \neq -1)$ return $M[i,j]$
3. if $(X[i] == Y[j])$
4. $M[i,j] = \text{LCS-recur}(X, Y, M, i-1, j-1) + 1$
5. else
6. $M[i,j] = \max(\text{LCS-recur}(X, Y, M, i-1, j), \text{LCS-recur}(X, Y, M, i, j-1))$
7. return $M[i,j]$

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3. Solve in a Good Order

$$LCS(i, j) = \begin{cases} 0 & \text{if } i = 0 \text{ or } j = 0 \\ LCS(i - 1, j - 1) + 1 & \text{if } X[i] = Y[j] \\ \max(LCS(i, j - 1), LCS(i - 1, j)) & \text{otherwise} \end{cases}$$

Y =									
		0	A	T	C	T	G	A	T
X =	0	0	0	0	0	0	0	0	0
	T	1	0	0	1	1	1	1	1
	G	2	0	0	1	1	1	2	2
	C	3	0	0	1	2	2	2	2
	A	4	0	1	1	2	2	2	3
	T	5	0	1	2	2	3	3	3
	A	6	0	1	2	2	3	3	4

To fill in cell (i, j) we need cells $(i - 1, j - 1)$, $(i - 1, j)$, $(i, j - 1)$
 Fill from Top->Bottom, Left->Right (with any preference)

LCS Length Algorithm

LCS-Length(X, Y) // Y for M's rows, X for its columns

1. $n = \text{length}(X)$ // get the # of symbols in X
2. $m = \text{length}(Y)$ // get the # of symbols in Y
3. for $i = 1$ to n $M[i,0] = 0$ // special case: X_0
4. for $j = 1$ to m $M[0,j] = 0$ // special case: Y_0
5. for $i = 1$ to n // for all X_i
6. for $j = 1$ to m // for all Y_j
7. if ($X[i] == Y[j]$)
8. $M[i,j] = M[i-1,j-1] + 1$
9. else $M[i,j] = \max(M[i-1,j], M[i,j-1])$
10. return $M[n,m]$ // return LCS length for Y and X

Run Time?

$$LCS(i, j) = \begin{cases} 0 & \text{if } i = 0 \text{ or } j = 0 \\ LCS(i - 1, j - 1) + 1 & \text{if } X[i] = Y[j] \\ \max(LCS(i, j - 1), LCS(i - 1, j)) & \text{otherwise} \end{cases}$$

		Y =							
		0	A	T	C	T	G	A	T
X =	0	0	0	0	0	0	0	0	0
	T	1	0	0	1	1	1	1	1
	G	2	0	0	1	1	1	2	2
	C	3	0	0	1	2	2	2	2
	A	4	0	1	1	2	2	2	3
	T	5	0	1	2	2	3	3	3
	A	6	0	1	2	2	3	3	4

Run Time: $\Theta(n \cdot m)$ (for $|X| = n, |Y| = m$)

Reconstructing the LCS

$$LCS(i, j) = \begin{cases} 0 & \text{if } i = 0 \text{ or } j = 0 \\ LCS(i - 1, j - 1) + 1 & \text{if } X[i] = Y[j] \\ \max(LCS(i, j - 1), LCS(i - 1, j)) & \text{otherwise} \end{cases}$$

Y =									
		0	A	T	C	T	G	A	T
X =	0	0	0	0	0	0	0	0	0
	T	0	0	1	1	1	1	1	1
	G	0	0	1	1	1	2	2	2
	C	0	0	1	2	2	2	2	2
	A	0	1	1	2	2	2	3	3
	T	0	1	2	2	3	3	3	4
	A	0	1	2	2	3	3	4	4

Start from bottom right,
 if symbols matched, print that symbol then go diagonally
 else go to largest adjacent

Reconstructing the LCS

$$LCS(i, j) = \begin{cases} 0 & \text{if } i = 0 \text{ or } j = 0 \\ LCS(i - 1, j - 1) + 1 & \text{if } X[i] = Y[j] \\ \max(LCS(i, j - 1), LCS(i - 1, j)) & \text{otherwise} \end{cases}$$

Y =

	0	A	T	C	T	G	A	T
X = 0	0	0	0	0	0	0	0	0
T	0	0	1	1	1	1	1	1
G	0	0	1	1	1	2	2	2
C	0	0	1	2	2	2	2	2
A	0	1	1	2	2	2	3	3
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Y =

	0	A	T	C	T	G	A	T
X = 0	0	0	0	0	0	0	0	0
T 1	0	0	1	1	1	1	1	1
G 2	0	0	1	1	1	2	2	2
C 3	0	0	1	2	2	2	2	2
A 4	0	1	1	2	2	2	3	3
T 5	0	1	2	2	3	3	3	4
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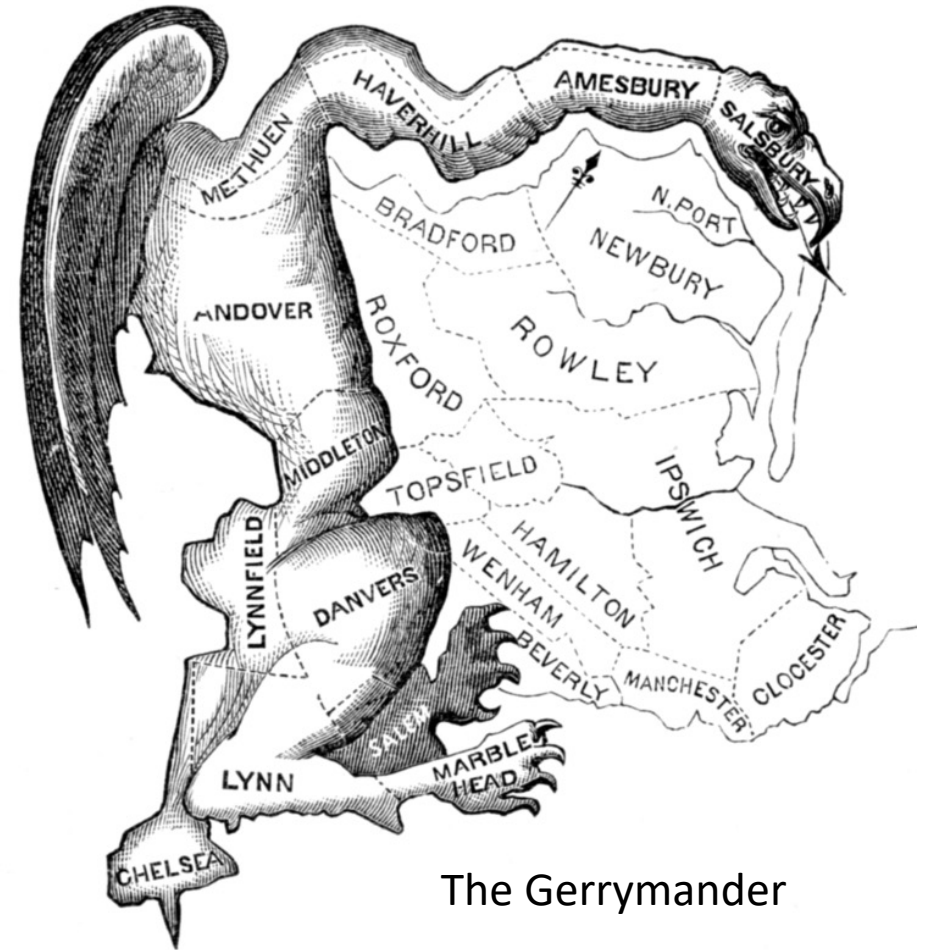
Supreme Court Associate Justice Anthony Kennedy gave no sign that he has abandoned his view that extreme partisan gerrymandering might violate the Constitution. | Eric Thayer/Getty Images

Supreme Court eyes partisan gerrymandering

Anthony Kennedy is seen as the swing vote that could blunt GOP's map-drawing successes.

Gerrymandering

- Manipulating electoral district boundaries to favor one political party over others
- Coined in an 1812 Political cartoon
- Governor **Elbridge Gerry** signed a bill that redistricted Massachusetts to benefit his Democratic-Republican Party



The Gerry-mander

According to the Supreme Court

- Gerrymandering cannot be used to:
 - Disadvantage racial/ethnic/religious groups
- It can be used to:
 - Disadvantage political parties

SUPREME COURT OF THE UNITED STATES

Syllabus

VIRGINIA HOUSE OF DELEGATES ET AL. *v.*
BETHUNE-HILL ET AL.

APPEAL FROM THE UNITED STATES DISTRICT COURT FOR THE
EASTERN DISTRICT OF VIRGINIA

No. 18–281. Argued March 18, 2019—Decided June 17, 2019

After the 2010 census, Virginia redrew legislative districts for the State’s Senate and House of Delegates. Voters in 12 impacted House districts sued two state agencies and four election officials (collectively, State Defendants), charging that the redrawn districts were racially gerrymandered in violation of the Fourteenth Amendment’s Equal Protection Clause. The House of Delegates and its Speaker (collectively, the House) intervened as defendants, participating in the bench trial, on appeal to this Court, and at a second bench trial, where a three-judge District Court held that 11 of the districts were unconstitutionally drawn, enjoined Virginia from conducting elections for those districts before adoption of a new plan, and gave the General Assembly several months to adopt that plan. Virginia’s Attorney General announced that the State would not pursue an appeal to this Court. The House, however, did file an appeal.

Held: The House lacks standing, either to represent the State’s interests or in its own right. Pp. 3–12.

SUPREME COURT OF THE UNITED STATES

Syllabus

RUCHO ET AL. *v.* COMMON CAUSE ET AL.

APPEAL FROM THE UNITED STATES DISTRICT COURT FOR THE
MIDDLE DISTRICT OF NORTH CAROLINA

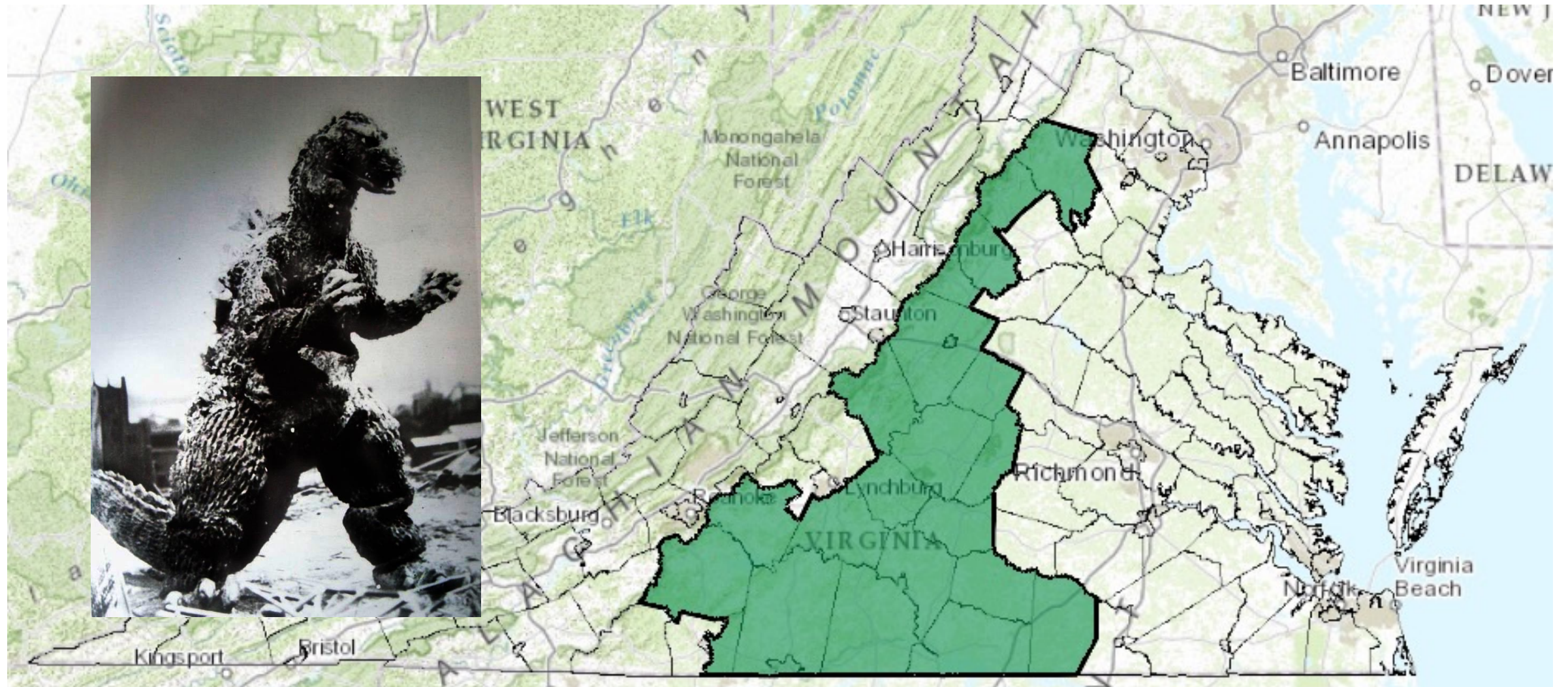
No. 18–422. Argued March 26, 2019—Decided June 27, 2019*

Voters and other plaintiffs in North Carolina and Maryland filed suits challenging their States’ congressional districting maps as unconstitutional partisan gerrymanders. The North Carolina plaintiffs claimed that the State’s districting plan discriminated against Democrats, while the Maryland plaintiffs claimed that their State’s plan discriminated against Republicans. The plaintiffs alleged violations of the First Amendment, the Equal Protection Clause of the Fourteenth Amendment, the Elections Clause, and Article 1, §2. The District Courts in both cases ruled in favor of the plaintiffs, and the defendants appealed directly to this Court.



Held: Partisan gerrymandering claims present political questions beyond the reach of the federal courts. Pp. 6–34.

(a) In these cases, the Court is asked to decide an important question of constitutional law. Before it does so, the Court “must find that the question is presented in a ‘case’ or ‘controversy’ that is . . . ‘of a Judiciary Nature.’” *DaimlerChrysler Corp. v. Cuno*, 547 U. S. 332, 342. While it is “the province and duty of the judicial department to . . . say what the law is,” *Marbury v. Madison*, 5 U. S. 137, 177.

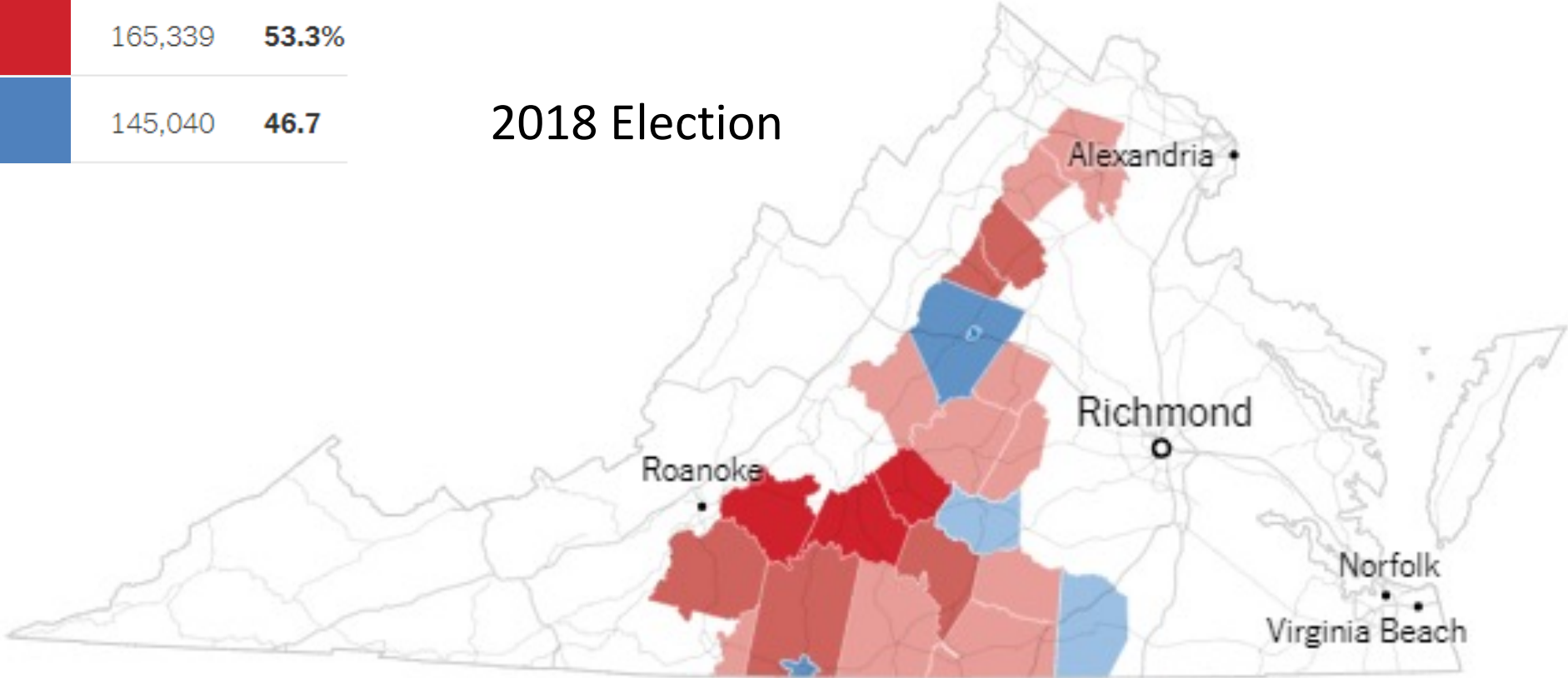
VA 5th District



VA 5th District

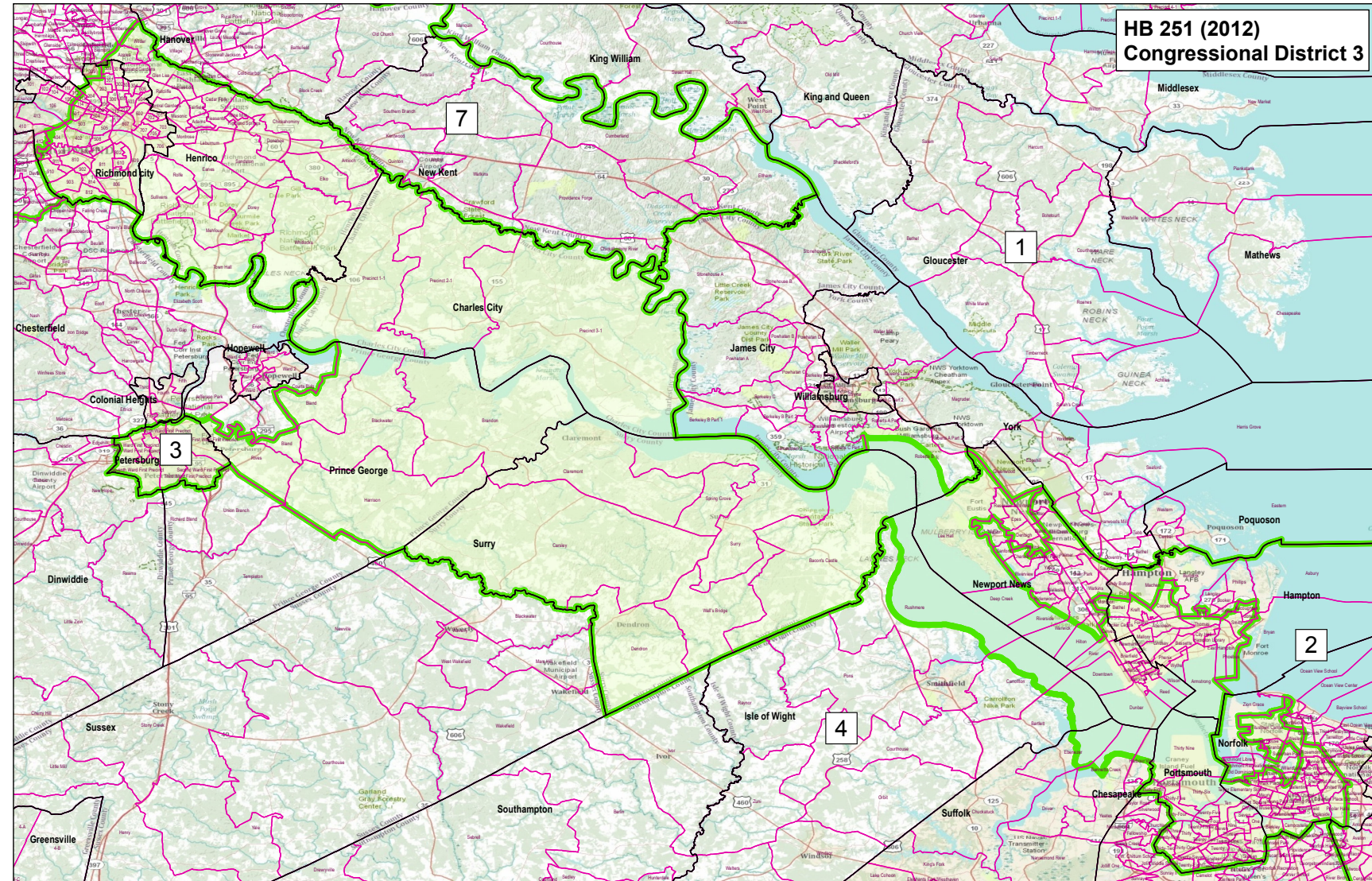
	Votes	Pct.
	165,339	53.3%
	145,040	46.7

2018 Election



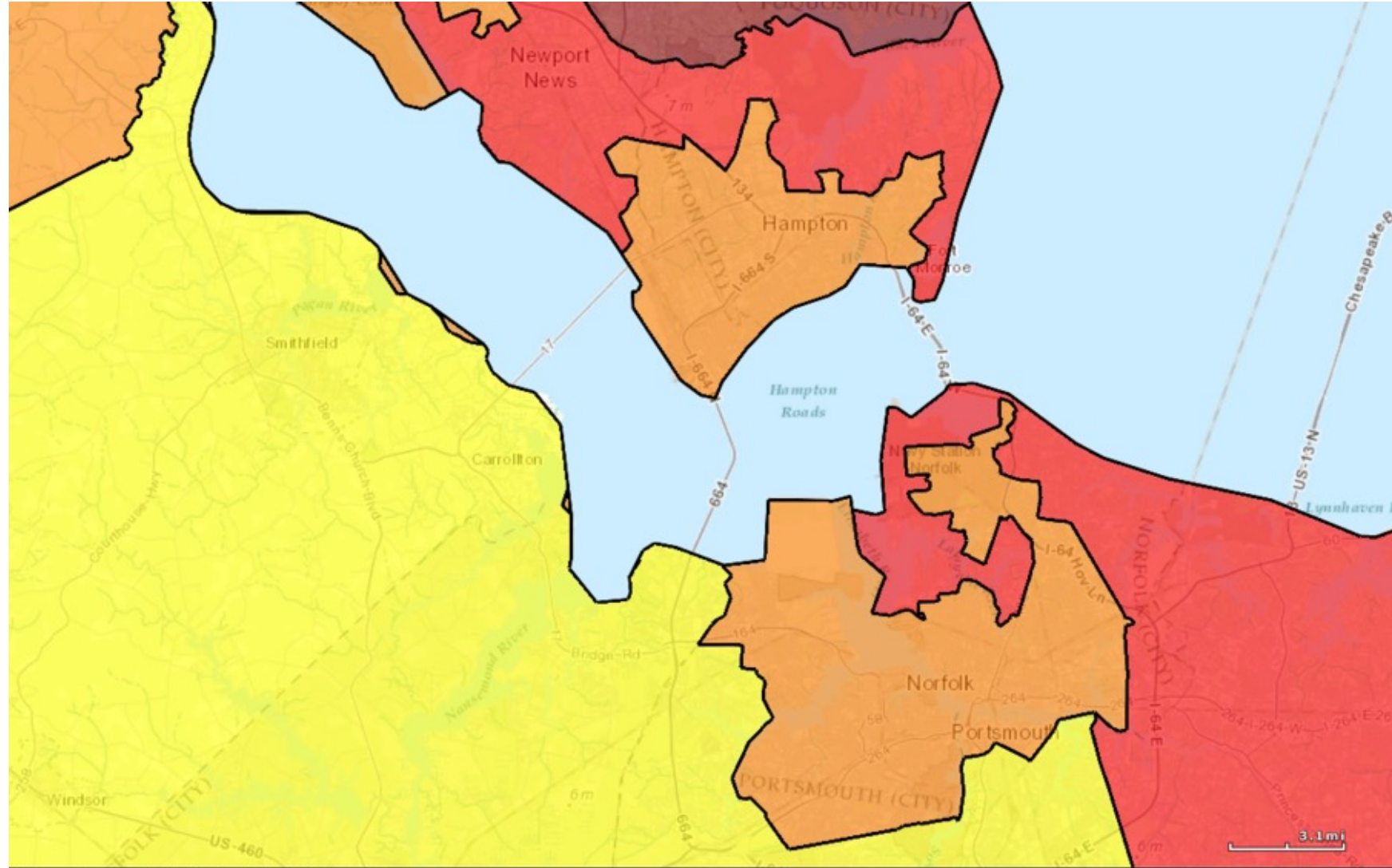
Gerrymandering Today

- Computers make it really effective



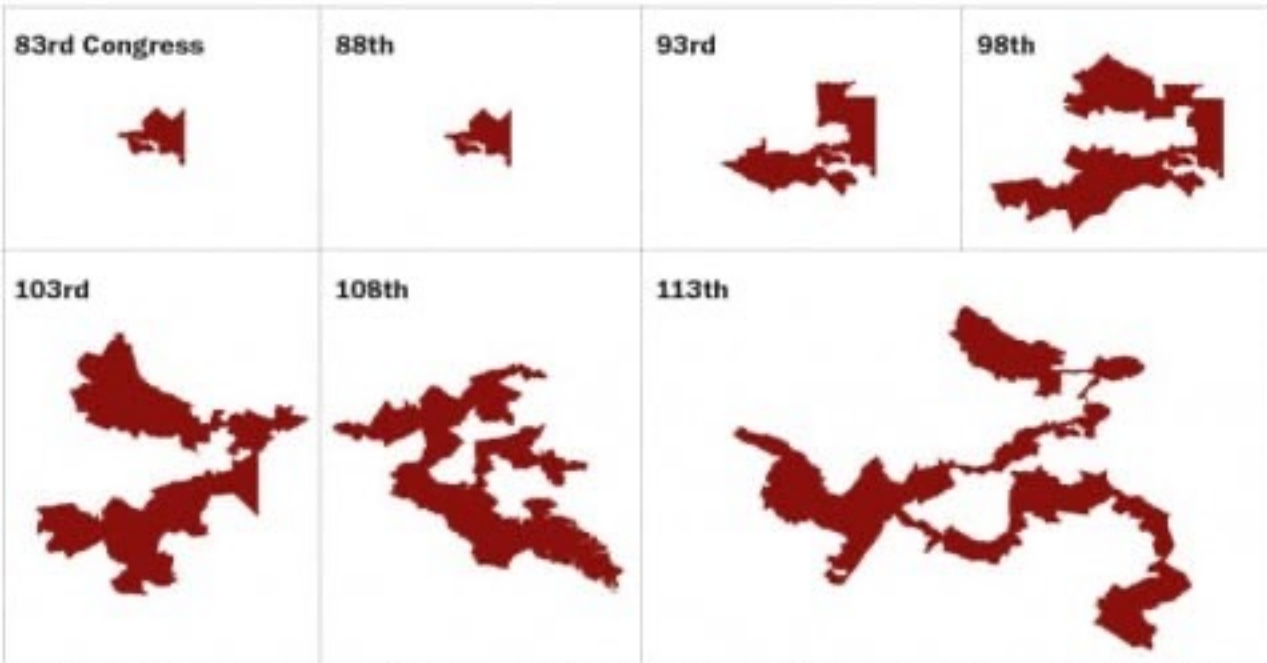
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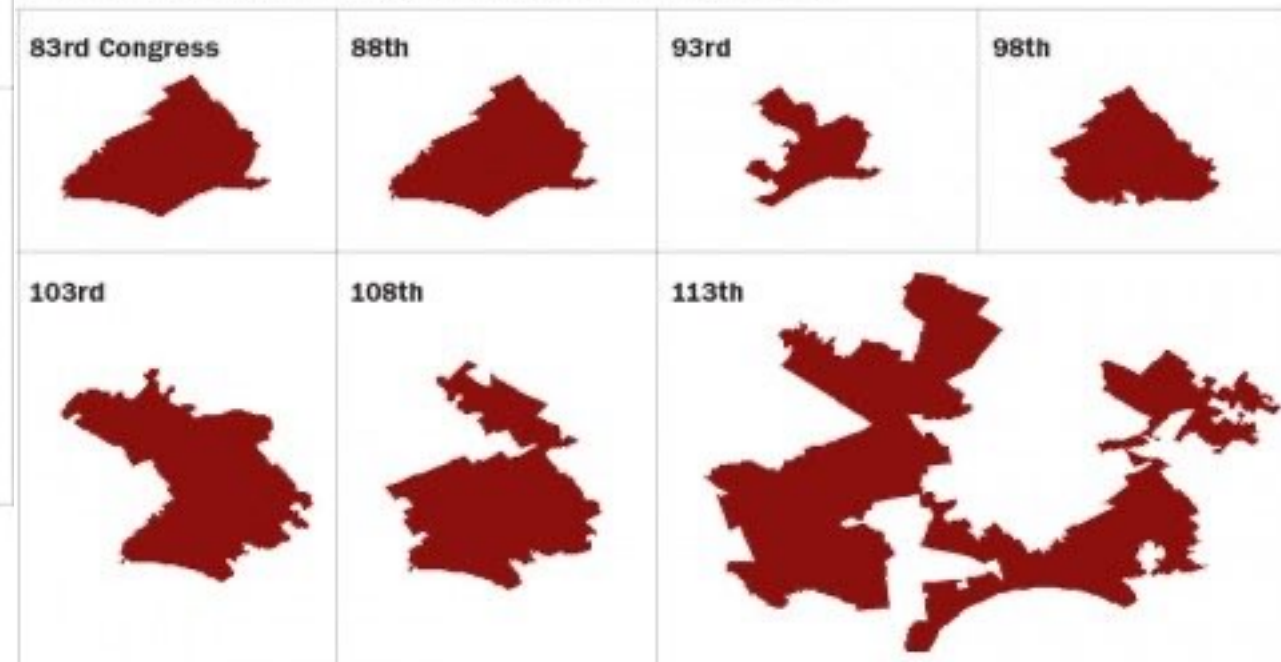
Gerrymandering Today

THE EVOLUTION OF MARYLAND'S THIRD DISTRICT



SOURCE: Shapefiles maintained by Jeffrey B. Lewis, Brandon DeVine, Lincoln Pritcher and Kenneth C. Martis, UCLA. Drawn to scale.
GRAPHIC: The Washington Post. Published May 20, 2014

THE EVOLUTION OF PENNSYLVANIA'S SEVENTH DISTRICT



SOURCE: Shapefiles maintained by Jeffrey B. Lewis, Brandon DeVine, Lincoln Pritcher and Kenneth C. Martis, UCLA. Drawn to scale.
GRAPHIC: The Washington Post. Published May 20, 2014

How does it work?

- States are broken into precincts
- All precincts have the same size
- We know voting preferences of each precinct
- Group precincts into districts to maximize the number of districts won by my party

Overall: R:217 D:183

R:65 D:35	R:45 D:55
R:60 D:40	R:47 D:53



VS



How does it work?

- States are broken into precincts
- All precincts have the same size
- We know voting preferences of each precinct
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Overall: R:217 D:183

R:65 D:35	R:45 D:55
R:60 D:40	R:47 D:53

R:125

R:92

R:65 D:35	R:45 D:55
R:60 D:40	R:47 D:53

R:112

R:105

R:65 D:35	R:45 D:55
R:60 D:40	R:47 D:53

Gerrymandering Problem Statement

- Given:
 - A list of precincts: p_1, p_2, \dots, p_n
 - Each containing m voters
- Output:
 - Districts $D_1, D_2 \subset \{p_1, p_2, \dots, p_n\}$
 - Where $|D_1| = |D_2|$
 - $R(D_1) > \frac{mn}{4}$ and $R(D_2) > \frac{mn}{4}$
 - $R(D_i)$ gives number of “Regular Party” voters in D_i
 - $R(D_i) > \frac{mn}{4}$ means D_i is majority “Regular Party”
 - “failure” if no such solution is possible

Valid Gerrymandering!

$$m \cdot \frac{n}{2} \cdot \frac{1}{2}$$

Dynamic Programming

- Requires **Optimal Substructure**
 - Solution to larger problem contains the solutions to smaller ones
- Idea:
 1. Identify the recursive structure of the problem
 - What is the “last thing” done?
 2. Save the solution to each subproblem in memory
 3. Select a good order for solving subproblems
 - “Top Down”: Solve each recursively
 - “Bottom Up”: Iteratively solve smallest to largest

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Consider the last precinct

After assigning the first $n - 1$ precincts

p_1, p_2, \dots, p_{n-1}

D_1
 k precincts
 x voters for R

D_2
 $n - k - 1$ precincts
 y voters for R

If we assign p_n to D_1

p_n

If we assign p_n to D_2

D_1
 $k + 1$ precincts
 $x + R(p_n)$ voters for R

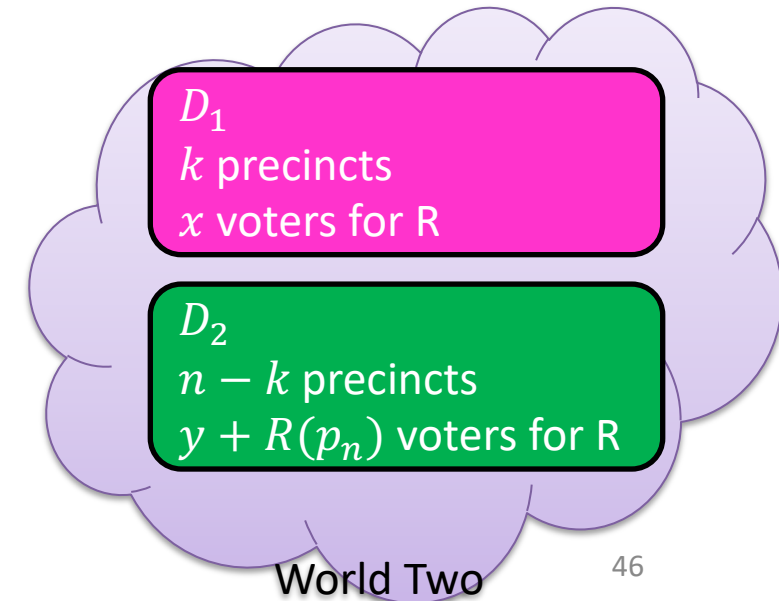
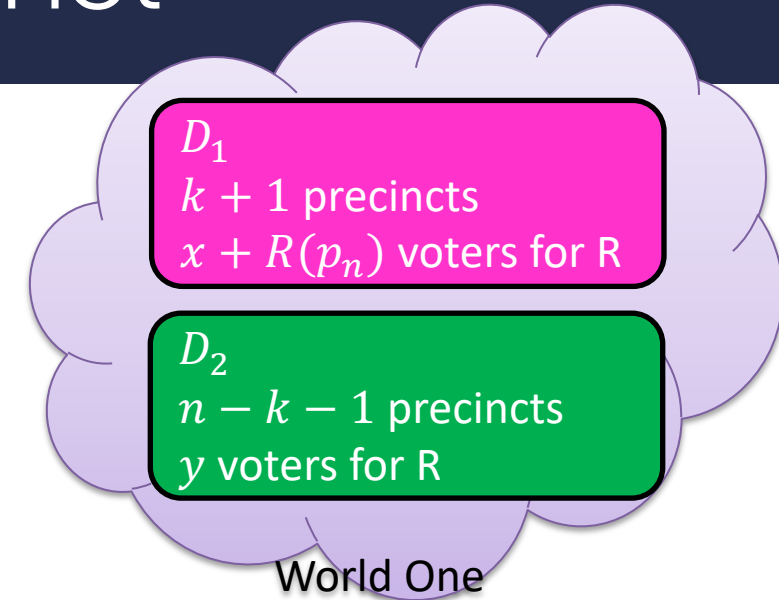
Valid gerrymandering if:

$$k + 1 = \frac{n}{2},$$
$$x + R(p_n), y > \frac{mn}{4}$$

D_2
 $n - k$ precincts
 $y + R(p_n)$ voters for R

Valid gerrymandering if:

$$n - k = \frac{n}{2},$$
$$x, y + R(p_n) > \frac{mn}{4}$$



Define Recursive Structure

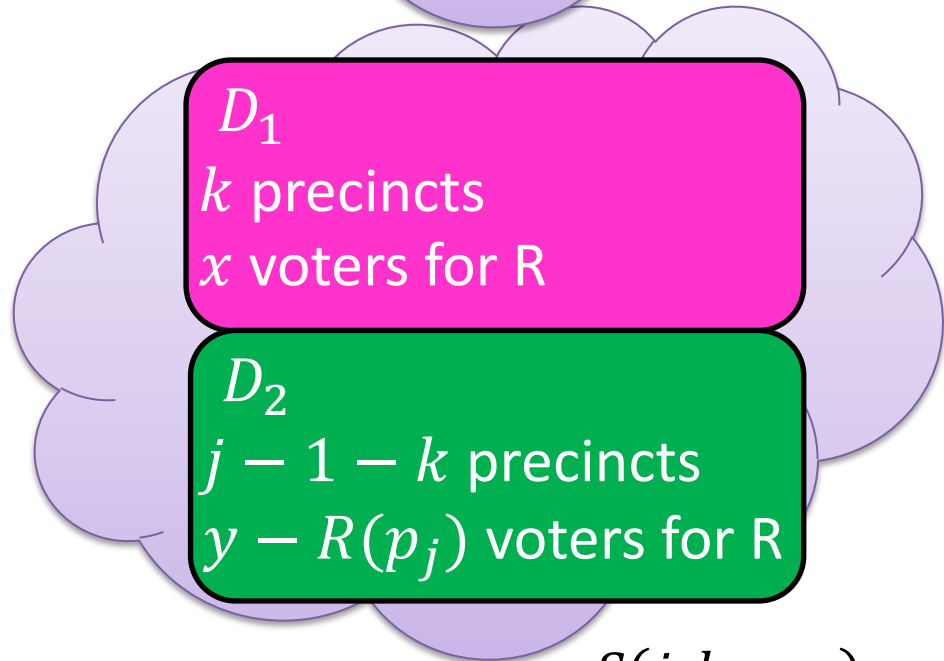
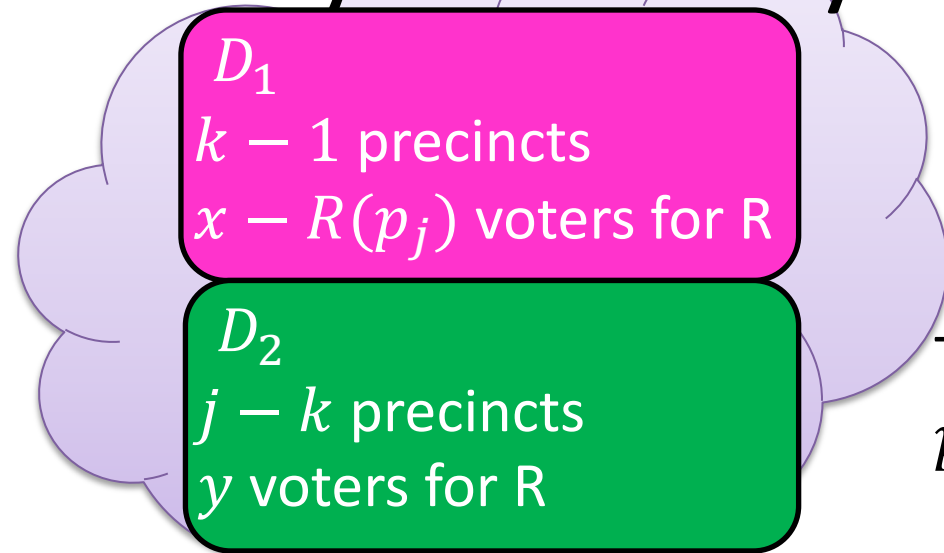
$S(j, k, x, y) =$ True if from among the first j precincts:
 k are assigned to D_1
exactly x vote for R in D_1
exactly y vote for R in D_2

$n \times n \times mn \times mn$

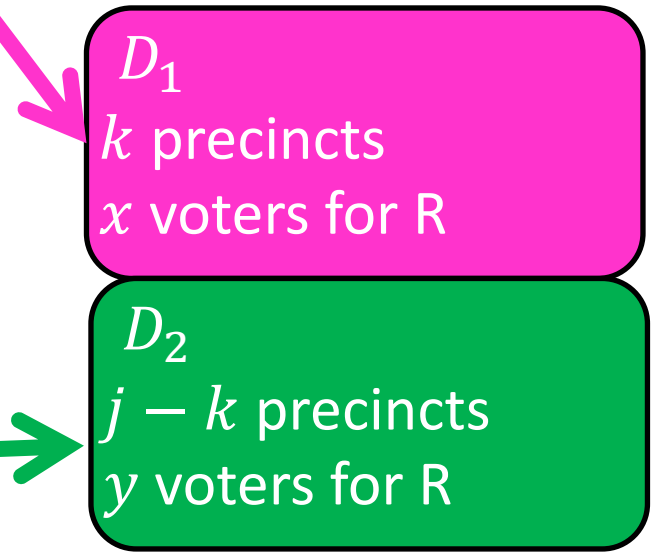
4D Dynamic Programming!!!

Two ways to satisfy $S(j, k, x, y)$:

$S(j, k, x, y) = \text{True}$ if:
from among the first j precincts
 k are assigned to D_1
exactly x vote for R in D_1
exactly y vote for R in D_2



OR



$$S(j, k, x, y) = S(j - 1, k - 1, x - R(p_j), y) \vee S(j - 1, k, x, y - R(p_j))$$

Final Algorithm

$$S(j, k, x, y) = S(j - 1, k - 1, x - R(p_j), y) \vee S(j - 1, k, x, y - R(p_j))$$

Initialize $S(0,0,0,0) = \text{True}$

for $j = 1, \dots, n$:

for $k = 1, \dots, \min(j, \frac{n}{2})$:

for $x = 0, \dots, jm$:

for $y = 0, \dots, jm$:

$S(j, k, x, y) =$

$$S(j - 1, k - 1, x - R(p_j), y) \vee S(j - 1, k, x, y - R(p_j))$$

Search for True entry at $S(n, \frac{n}{2}, > \frac{mn}{4}, > \frac{mn}{4})$

$S(j, k, x, y) = \text{True}$ if:

from among the first j precincts

k are assigned to D_1

exactly x vote for R in D_1

exactly y vote for R in D_2

Run Time

$$S(j, k, x, y) = S(j - 1, k - 1, x - R(p_j), y) \vee S(j - 1, k, x, y - R(p_j))$$

Initialize $S(0,0,0,0) = \text{True}$

n for $j = 1, \dots, n$:

$\frac{n}{2}$ for $k = 1, \dots, \min(j, \frac{n}{2})$:

nm for $x = 0, \dots, jm$:

nm for $y = 0, \dots, jm$:

$S(j, k, x, y) =$

$$S(j - 1, k - 1, x - R(p_j), y) \vee S(j - 1, k, x, y - R(p_j))$$

Search for True entry at $S(n, \frac{n}{2}, > \frac{mn}{4}, > \frac{mn}{4})$

$\Theta(n^4 m^2)$

$$\Theta(n^4 m^2)$$

- Input: list of precincts (size n), number of voters (integer m)
- Runtime depends on the *value of m* , not *size of m*
 - Run time is exponential in *size* of input
 - Input size is $n + |m| = n + \log m$
- Note: Gerrymandering is NP-Complete