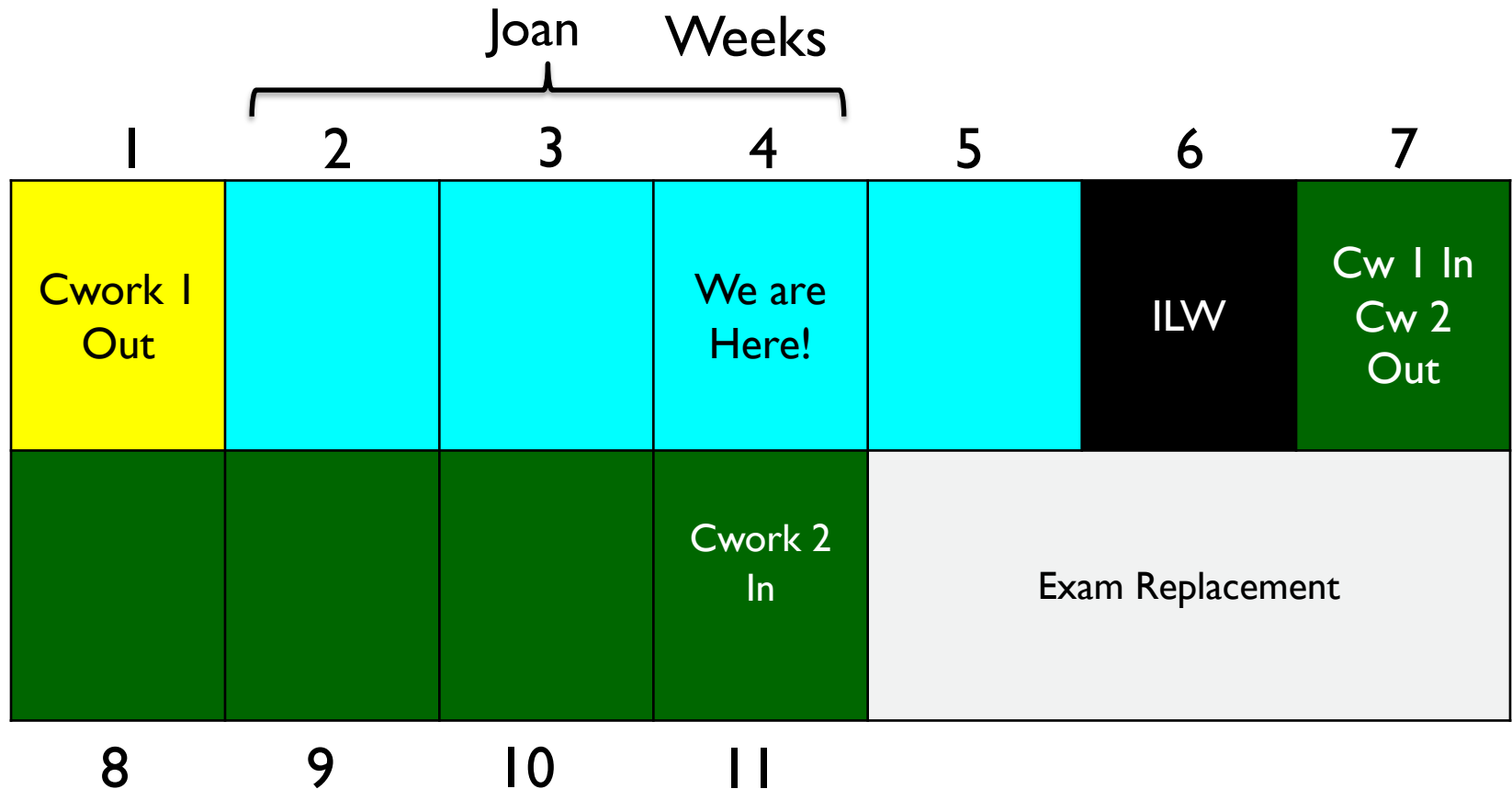


# CS4402: Constraint Programming

Week 4, Lecture 1: Symmetry

Check MMS  
for actual c/w  
deadlines

# Roadmap



Basics

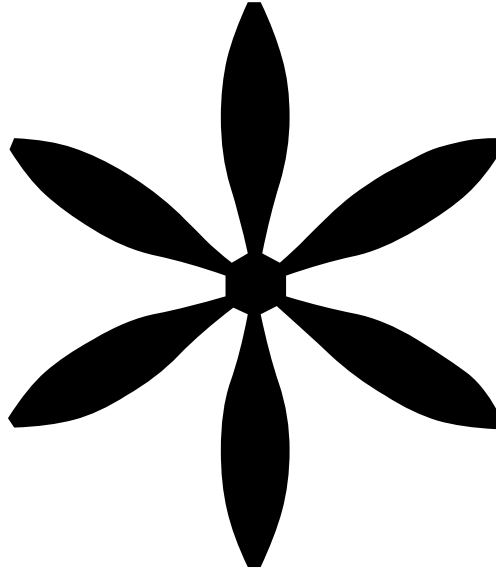
Modelling

Search and Propagation

# Symmetry in Constraint Models

# Symmetry

- A familiar everyday concept:



- A **structure-preserving transformation.**

# Symmetry in Constraint Models

- A symmetry can be characterised as a **bijection** on assignments.
  - i.e. a one-to-one mapping.
- Partitions complete assignments into **equivalence classes**.
  - In every class, all members are solutions, or no member is.
- That is, symmetry **preserves solutionhood**.

# Symmetry in Constraint Models

- Why do we care?
- Frequently present in the structure of a CSP.
- If we don't deal with it, can lead to a lot of wasted effort for systematic search.

# Two Common Special Cases

- **Variable** symmetry.
  - Bijection can be characterised in terms of variables alone.
- **Value** symmetry.
  - Bijection can be characterised in terms of values alone.

# Variable Symmetry



# Variable Symmetry Example

	A Solution			
q1		Q		
q2				Q
q3	Q			
q4			Q	

Flip Horizontal

	A Solution			
q4		Q		
q3				Q
q2	Q			
q1			Q	

Equivalently:

	A Solution			
q1			Q	
q2	Q			
q3				Q
q4		Q		

- We can express this symmetry in terms of a bijection on the **variables**:
  - $q1 \rightarrow q4$ ,
  - $q2 \rightarrow q3$ ,
  - $q3 \rightarrow q2$ ,
  - $q4 \rightarrow q1$
- A bijection on the **values** would capture the transformation of this particular solution – but not the **general** horizontal flip.

# Variable Symmetry Example

	Non-solution			
q1		Q		
q2				Q
q3	Q			
q4		Q		

Flip Horizontal

	Non-solution			
q4		Q		
q3				Q
q2	Q			
q1		Q		

Equivalently

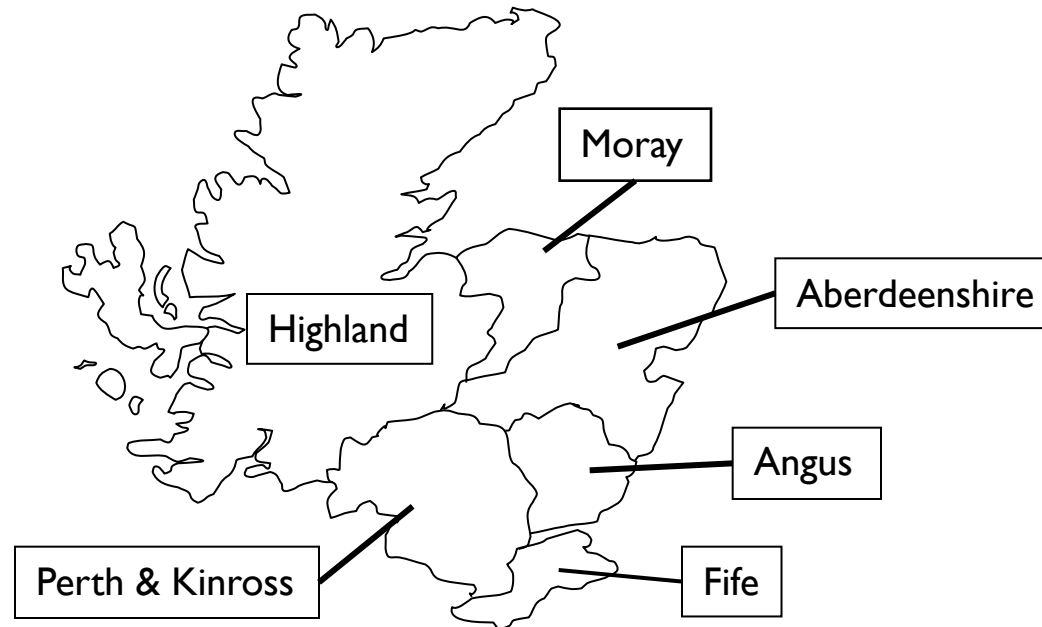
	Non-solution			
q1		Q		
q2	Q			
q3				Q
q4		Q		

- We can express this symmetry in terms of a bijection on the **variables**:

- $q1 \rightarrow q4$ ,
- $q2 \rightarrow q3$ ,
- $q3 \rightarrow q2$ ,
- $q4 \rightarrow q1$

# Value Symmetry

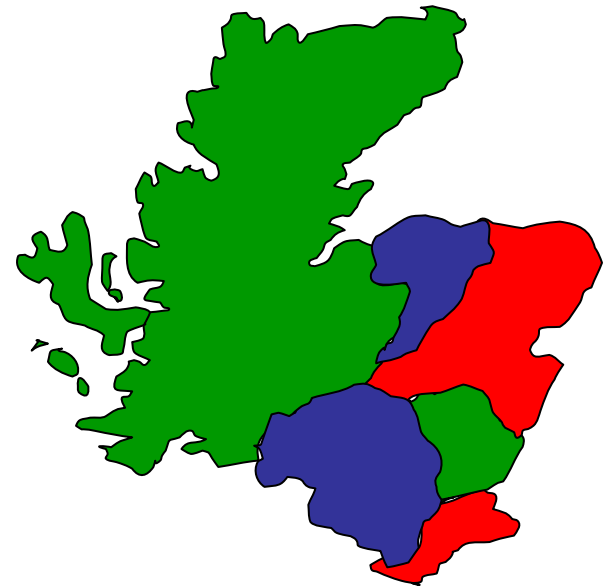
# A Graph Colouring Problem



- We want to colour this map with the colours red green and blue such that adjacent regions are not coloured the same.

# A Solution

- Highland = green
- Moray = blue
- Aberdeenshire = red
- Angus = green
- Fife = red
- Perth and Kinross = blue



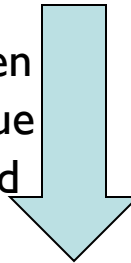
Can you see the symmetry?

# Symmetric Solutions

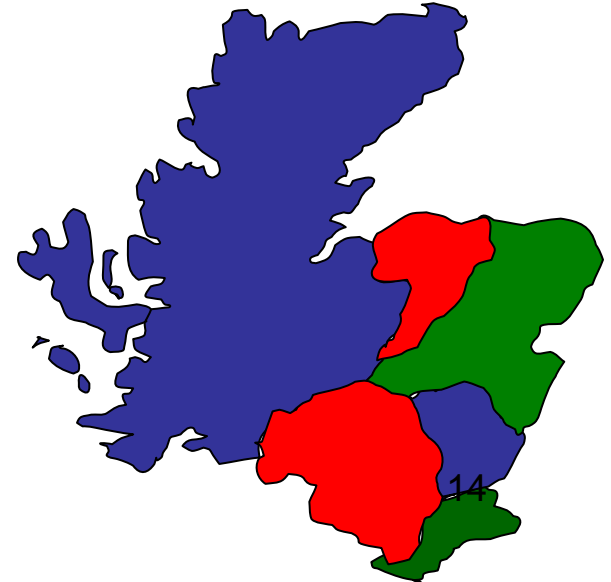
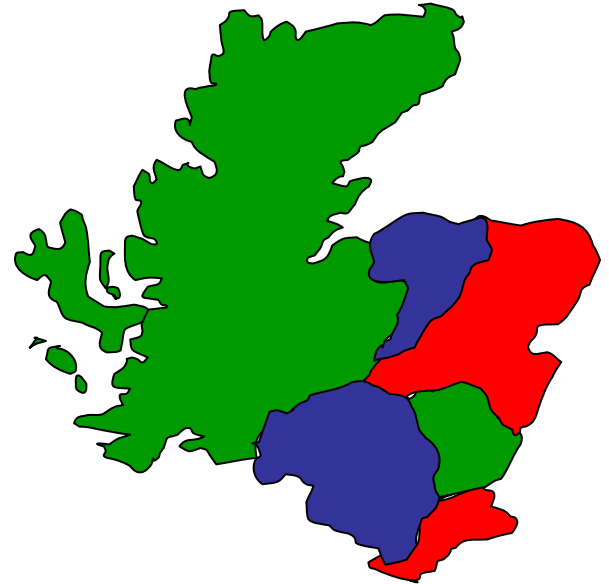
- Highland = green
- Moray = blue
- Aberdeenshire = red
- Angus = green
- Fife = red
- Perth and Kinross = blue

We've  
permuted the  
**values.**

Red → green  
green → blue  
blue → red



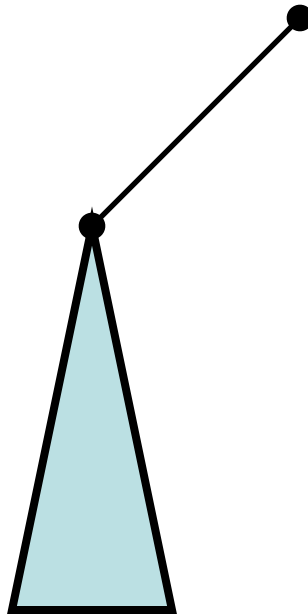
- Highland = blue
- Moray = red
- Aberdeenshire = green
- Angus = blue
- Fife = green
- Perth and Kinross = red



# Symmetry: Consequences

# What are The Consequences of a Model Containing Symmetry?

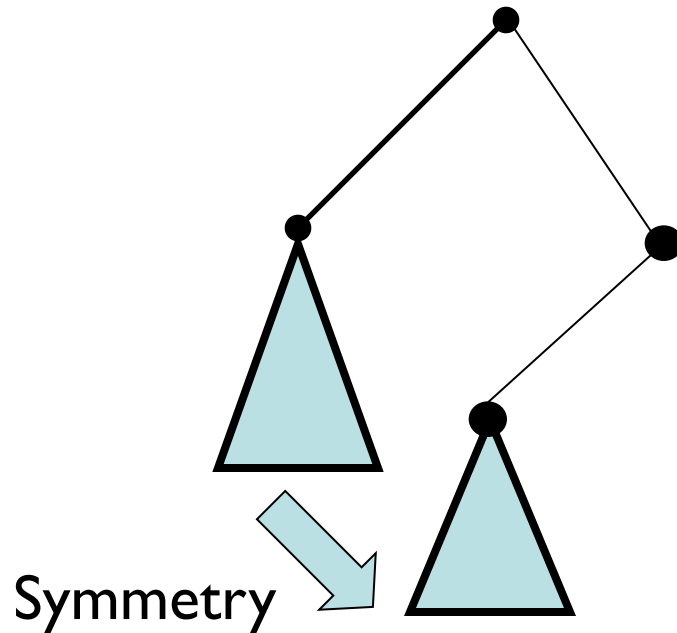
- Assume we have explored a sub-tree in which there are no solutions:





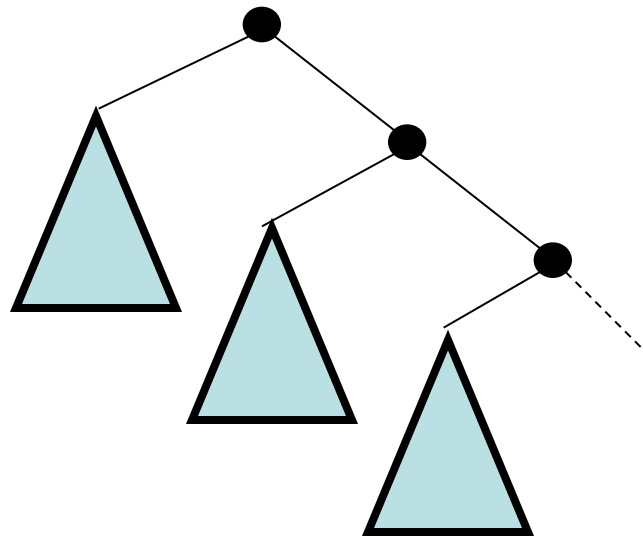
# What are The Consequences of a Model Containing Symmetry?

- Symmetries in the model can map this fruitless subtree into another:



# What are The Consequences of a Model Containing Symmetry?

- If we don't deal with the symmetry, a systematic search will explore all symmetric variants of this sub-tree:



# Symmetry: Origins

# How Does Symmetry Arise?

- Symmetry enters a model from 2 sources:
  - I. It is inherent in the problem:

	A Solution			
q1		Q		
q2				Q
q3	Q			
q4			Q	

By its nature, the  $n$ -queens problem has the symmetries of the square

# How Does Symmetry Arise?

- Symmetry enters a model from 2 sources:
  1. It is introduced during the modelling process.
- Remember the explicit model of sets:

1	2	3	4	...	n-1	n
0..m	0..m	0..m	0..m		0..m	0..m

AllDifferent

- Now there is a 1<sup>st</sup> element, a 2<sup>nd</sup> element, etc...

# How Does Symmetry Arise?

- Given any (non-)solution:

1	2	3	4
a	b	c	d

- Can permute the elements to obtain a (non-)solution:

1	2	3	4
b	d	c	a

NB This is again a variable symmetry: we have given arbitrarily named indices to the elements of the set.

**NB Knowing how symmetry is introduced during modelling can make some symmetry detection easy.**

# Symmetry-breaking Constraints

# How Can Adding Constraints Affect Symmetry?

- By disallowing some of the elements of an **equivalence class** of assignments.

M	1	2	3	4	...	n-1	n
	0..m	0..m	0..m	0..m		0..m	0..m

AllDifferent(M)

- Consider ordering the elements of M:

$$M[1] \leq M[2] \leq M[3] \leq \dots \leq M[n-1] \leq M[n]$$

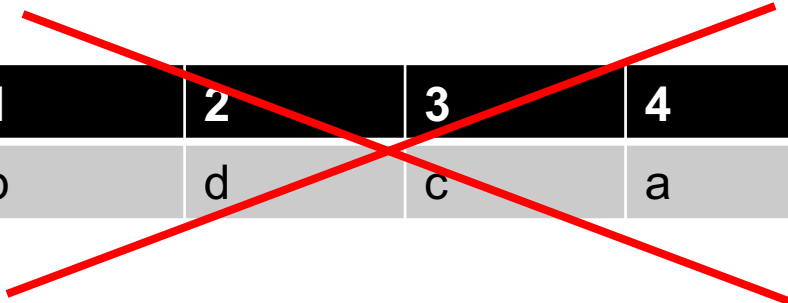


# How Can Adding Constraints Affect Symmetry?

- Ordering the elements distinguishes one element in each equivalence class of assignments:

1	2	3	4
a	b	c	d

1	2	3	4
b	d	c	a



- If we can propagate symmetry-breaking constraints effectively, they can prune the search tree drastically.

# Valid Symmetry-breaking Constraints: Properties

- Leave at least one element in each equivalence class of assignments.
- So they **throw away solutions** to the original problem?
  - Yes: but we can recover those solutions by applying the symmetry to the solutions allowed.
- If we add enough symmetry-breaking constraints we can break all symmetry.

# How do we Formulate Symmetry-breaking Constraints?

- Define a canonical solution and add constraints to choose it.
- Need:
  - An ordering on **variables**.
  - An ordering on **solutions**.

# Orderings on Variables

- Simple: we just choose one.

<b>M</b>	<b>1</b>	<b>2</b>	<b>3</b>	<b>4</b>	<b>...</b>	<b>n-1</b>	<b>n</b>
	0..m	0..m	0..m	0..m		0..m	0..m

- Here, the natural thing to do is choose an ordering along the indices

# Orderings on the Solutions

- Use a **lexicographic** ordering:

v1	1	2	3	4
	a	b	c	d

$$\mathbf{v}_1 \leq_{\text{lex}} \mathbf{v}_2$$

v2	1	2	3	4
	b	d	c	a

# The Lex-leader Method

- Choose an ordering on the variables.
- Add one lexicographic ordering constraint per symmetry.
- Good news: this breaks all symmetry.
- Bad news: you might need a huge number of such constraints
  - But sometimes the lex constraints collapse into a simpler form (e.g. the explicit set representation).
- Compromise: use a subset of the lex-leader constraints.

# Lex Leader Example

A	B	C
D	E	F

- Assume that this matrix has row and column symmetry.
- Choose ordering on variables ABCDEF

# Lex Leader Example

A	B	C
D	E	F

- ABCDEF  $\leq_{\text{lex}}$  DEFABC (swap rows).
- ABCDEF  $\leq_{\text{lex}}$  ACBDFE (swap last two cols).
- ABCDEF  $\leq_{\text{lex}}$  DFEACB (swap rows & last 2 cols).
- ...
- We are preventing it from being possible to obtain a “smaller” assignment than the one we currently have by applying a symmetry.



# The BIBD has Row and Column Symmetry

- This is **very** common in matrix models.

0	0	0	0	1	1	1
0	0	1	1	0	0	1
0	1	0	1	0	1	0
0	1	1	0	1	0	0
1	0	0	1	1	0	0
1	0	1	0	0	1	0
1	1	0	0	0	0	1

0	0	1	0	1	0	1
0	0	0	1	0	1	1
0	1	0	0	1	1	0
0	1	1	1	0	0	0
1	0	0	1	1	0	0
1	0	1	0	0	1	0
1	1	0	0	0	0	1

# Lex-leader for BIBD

- Requires  **$b!v!$**  lex constraints.

Blocks

Objects

1	2	...				
					...	49

# Lex<sup>2</sup> ordering for Row/Column Symmetry

- A significant fraction of lex-leader for row/column symmetry can be represented by lex ordering the rows and columns.

0	0	0	0	1	1	1
0	0	1	1	0	0	1
0	1	0	1	0	1	0
0	1	1	0	1	0	0
1	0	0	1	1	0	0
1	0	1	0	0	1	0
1	1	0	0	0	0	1

0	0	1	0	1	0	1
0	0	0	1	0	1	1
0	1	0	0	1	1	0
0	1	1	1	0	0	0
1	0	0	1	1	0	0
1	0	1	0	0	1	0
1	1	0	0	0	0	1

# Simple example of Lex<sup>2</sup>

A	B	C
D	E	F

- i. **BE**  $\leq$  **CF**
- ii. **AD**  $\leq$  **BE**
- iii. **ABC**  $\leq$  **DEF**

# Lex Leader Method:

## On the same example

A	B	C
D	E	F

1.  $ABCDEF \leq ACBDFE$
2.  $ABCDEF \leq BCAEFD$
3.  $ABCDEF \leq BACEDF$
4.  $ABCDEF \leq CABFDE$
5.  $ABCDEF \leq CBAFED$
6.  $ABCDEF \leq DFEACB$
7.  $ABCDEF \leq EFDBCA$
8.  $ABCDEF \leq EDFBAC$
9.  $ABCDEF \leq FDECAB$
10.  $ABCDEF \leq FEDCBA$
11.  $ABCDEF \leq DEFABC$

# Lex Leader vs Lex<sup>2</sup>

A	B	C
D	E	F

i. **BE** ≤ **CF**

ii. **AD** ≤ **BE**

iii. **ABC** ≤ **DEF**

1. ABCDEF ≤ ACBDFE
2. ABCDEF ≤ BCAEFD
3. ABCDEF ≤ BACEDF
4. ABCDEF ≤ CABFDE
5. ABCDEF ≤ CBAFED
6. ABCDEF ≤ DFEACB

7. ABCDEF ≤ EFDBCA
8. ABCDEF ≤ EDFBAC
9. ABCDEF ≤ FDECAB
10. ABCDEF ≤ FEDCBA
11. ABCDEF ≤ DEFABC

# Lex Leader Method: Connection to Lex<sup>2</sup>

A	B	C
D	E	F

i. **BE** ≤ **CF**

ii. **AD** ≤ **BE**

iii. **ABC** ≤ **DEF**

1. **ABCDEF** ≤ **ACBDFE**

2. **ABCDEF** ≤ **BCAEFD**

3. **ABCDEF** ≤ **BACEDF**

4. **ABCDEF** ≤ **CABFDE**

5. **ABCDEF** ≤ **CBAFED**

6. **ABCDEF** ≤ **DFEACB**

7. **ABCDEF** ≤ **EFDBCA**

8. **ABCDEF** ≤ **EDFBAC**

9. **ABCDEF** ≤ **FDECAB**

10. **ABCDEF** ≤ **FEDCBA**

11. **ABCDEF** ≤ **DEFABC**

# Lex Leader Method: Connection to Lex<sup>2</sup>

i. **BE** ≤ **CF**

A	B	C
D	E	F

I. **ABCDEF** ≤ **ACBDFE**

- Assume **ABCDEF** ≤ **ACBDFE**
  - A always equals A and D always equals D
- So **BCEF** ≤ **CBFE**
  - If B = C then obviously C=B
- So **BEF** ≤ **CFE**
  - If E = F then obviously F=E
- So **BE** ≤ **CF**



# Lex Leader Method: Connection to Lex<sup>2</sup>

i. **BE** ≤ **CF**

A	B	C
D	E	F

I. **ABCDEF** ≤ **ACBDFE**

- These two constraints are actually *equivalent*.
- Similarly for the other constraints in Lex<sup>2</sup>
- But Lex Leader has additional constraints
- So Lex Leader sometimes breaks *more* symmetry than Lex<sup>2</sup>
- While the extra constraints means it can be much more expensive to reason with than Lex<sup>2</sup>