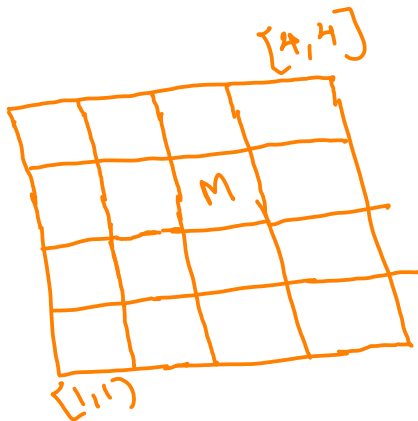


# Questions regarding assignment2



# Curiosity example

1. Query: Who killed the cat?

KB  $\models$  *Kills*(*x*, *Tuna*) ?

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$\neg \exists x \text{Kills}(x, \text{Tuna}) \equiv \forall x \neg \text{Kills}(x, \text{Tuna})$

- ▶ Different substitutions of  $x$  is allowed:

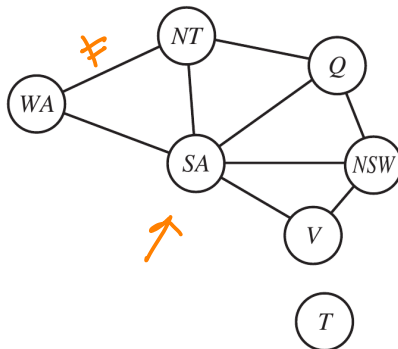




# Chapter 6: Constraint Satisfaction Problem



(a)



(b)

**Figure 6.1** (a) The principal states and territories of Australia. Coloring this map can be viewed as a constraint satisfaction problem (CSP). The goal is to assign colors to each region so that no neighboring regions have the same color. (b) The map-coloring problem represented as a constraint graph.

# Constraint Satisfaction Problem

- ▶ Australian map coloring problem

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- ▶ Australian map coloring problem
- ▶ Problem definition :  $X, D$  and  $C$

$\{\text{Red}, G, B\}$

↑   ↑   ↑

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  1.  $X = \{X_1, X_2, \dots, X_n\}$

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- ▶ Australian map coloring problem
- ▶ Problem definition :  $X, D$  and  $C$ 
  1.  $X = \{X_1, X_2, \dots, X_n\}$
  2.  $D = \{D_1, D_2, \dots, D_n\}$

# Constraint Satisfaction Problem

- ▶ Australian map coloring problem
- ▶ Problem definition :  $X$ ,  $D$  and  $C$ 
  1.  $X = \{X_1, X_2, \dots, X_n\}$
  2.  $D = \{D_1, D_2, \dots, D_n\}$
  3.  $C$  is a set of constraints.



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- ▶ Australian map coloring problem
- ▶ Problem definition :  $X, D$  and  $C$ 
  1.  $X = \{X_1, X_2, \dots, X_n\}$
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- ▶ Consistent assignment



# Constraint Satisfaction Problem

- ▶ Australian map coloring problem
- ▶ Problem definition :  $X, D$  and  $C$ 
  1.  $X = \{X_1, X_2, \dots, X_n\}$
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- ▶ Consistent assignment
- ▶ Complete assignment

# Constraint Satisfaction Problem

- ▶ Australian map coloring problem
- ▶ Problem definition :  $X, D$  and  $C$ 
  1.  $X = \{X_1, X_2, \dots, X_n\}$
  2.  $D = \{D_1, D_2, \dots, D_n\}$
  3.  $C$  is a set of constraints.
- ▶ Consistent assignment
- ▶ Complete assignment
- ▶ Solution : consistent and complete

# Constraint Satisfaction Problem (CSP)

- ▶ constraint graph vs. state space graph

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- ▶  $X = \{WA, NT, Q, NSW, V, SA, T\}$

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- ▶ constraint graph vs. state space graph
- ▶  $X = \{WA, NT, Q, NSW, V, SA, T\}$
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- ▶  $C = \{WA \neq NT, WA \neq SA, \dots, V \neq NSW\}$



# Constraint Satisfaction Problem (CSP)

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- ▶  $D_i = \{red, green, blue\}$
- ▶  $C = \{WA \neq NT, WA \neq SA, \dots, V \neq NSW\}$
- ▶ Multiple solutions are possible.

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- ▶ constraint graph vs. state space graph
- ▶  $X = \{WA, NT, Q, NSW, V, SA, T\}$
- ▶  $D_i = \{red, green, blue\}$
- ▶  $C = \{WA \neq NT, WA \neq SA, \dots, V \neq NSW\}$
- ▶ Multiple solutions are possible.
- ▶ Goal: understand general approaches for solving CSPs.

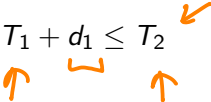


# Car assembly scheduling


# Car assembly scheduling

►  $X = \{Axle_F, Axle_B, \overset{\downarrow}{Wheel_{RF}}, \overset{\downarrow}{Wheel_{RB}}, Nuts_{RF}, Nuts_{RB},$   
 $\overset{\uparrow}{Cap_{RF}}, \overset{\uparrow}{Cap_{RB}}, \underset{\text{---}}{Inspect}\}$


# Car assembly scheduling

- ▶  $X = \{Axle_F, Axle_B, Wheel_{RF}, Wheel_{RB}, Nuts_{RF}, Nuts_{RB}, Cap_{RF}, Cap_{RB}, Inspect\}$
- ▶ Precedence constraints :  $T_1 + d_1 \leq T_2$ 

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  - ▶  $Axle_F + 10 \leq Wheel_{RF}; Axle_B + 10 \leq Wheel_{RB}$ 

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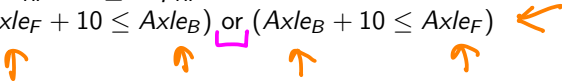
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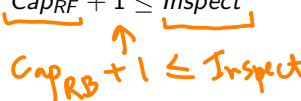
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  - ▶  $Nuts_{RF} + 2 \leq Cap_{RF}$ 



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  - ▶  $Wheel_{RF} + 1 \leq Nuts_{RF}; Wheel_{RB} + 1 \leq Nuts_{RB}$
  - ▶  $Nuts_{RF} + 2 \leq Cap_{RF}$
  - ▶  $(Axle_F + 10 \leq Axle_B) \text{ or } (Axle_B + 10 \leq Axle_F)$  ←

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  - ▶  $(Axle_F + 10 \leq Axle_B) \text{ or } (Axle_B + 10 \leq Axle_F)$
  - ▶  $Cap_{RF} + 1 \leq Inspect$   
  
 $Cap_{RB} + 1 \leq Inspect$



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  - ▶  $Nuts_{RF} + 2 \leq Cap_{RF}$
  - ▶  $(Axle_F + 10 \leq Axle_B) \text{ or } (Axle_B + 10 \leq Axle_F)$
  - ▶  $Cap_{RF} + 1 \leq Inspect$   

- ▶  $D_i = \{1, 2, 3, \dots, 27\}$

# Types of constraints

- ▶ Unary constraint

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$WA \neq \textit{Green}$



# Types of constraints

- ▶ Unary constraint  
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$WA \neq \textit{Green}$

- ▶ Binary constraint

$WA \neq NT$

# Types of constraints

- ▶ Unary constraint  
 $WA \neq \textit{Green}$
- ▶ Binary constraint  
 $WA \neq \textit{NT}$
- ▶ Global constraint

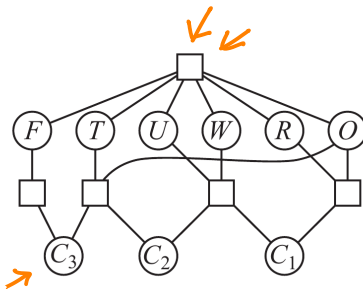
# Types of constraints

- ▶ Unary constraint  
 $WA \neq \textit{Green}$
- ▶ Binary constraint  
 $WA \neq \textit{NT}$
- ▶ Global constraint  
 $\textit{Alldiff}(A, B, C, D)$

# Cryptoarithmic Problem

$$\begin{array}{r} \phantom{00}TWO \\ + TWO \\ \hline FOUR \end{array}$$

Handwritten annotations:  $C_3, C_2, C_1$  above the first row, and arrows pointing to the first and second columns of the sum.

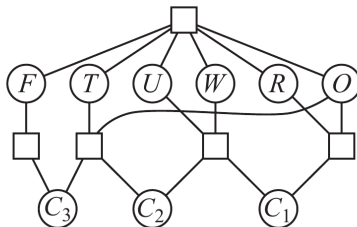


► Alldiff(F, T, U, W, R, O)



# Cryptoarithmic Problem

$$\begin{array}{r} T \ W \ O \\ + \ T \ W \ O \\ \hline F \ O \ U \ R \end{array}$$

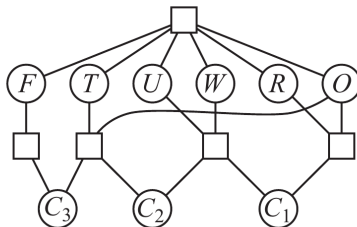


►  $Alldiff(F, T, U, W, R, O)$

It is given that all letters are distinct.


# Cryptoarithmic Problem

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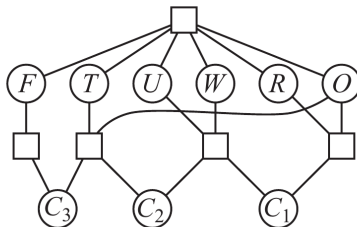
It is given that all letters are distinct.

- ▶  $O + O = 10 \times C_1 + R$  

# Cryptarithmic Problem

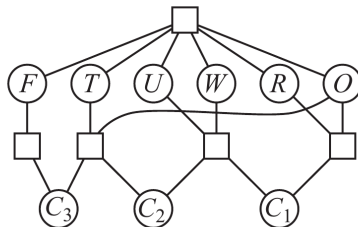
$$\begin{array}{r} T \ W \ O \\ + \ T \ W \ O \\ \hline F \ O \ U \ R \end{array}$$



- ▶  $Alldiff(F, T, U, W, R, O)$   
It is given that all letters are distinct.
- ▶  $O + O = 10 \times C_1 + R$
- ▶  $C_1 + W + W = 10 \times C_2 + U$

# Cryptoarithmic Problem

$$\begin{array}{r} T \ W \ O \\ + \ T \ W \ O \\ \hline F \ O \ U \ R \end{array}$$



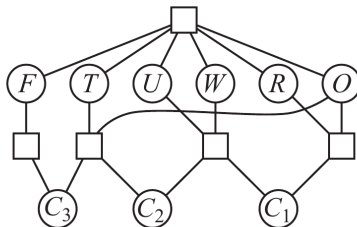
- ▶  $Alldiff(F, T, U, W, R, O)$

It is given that all letters are distinct.

- ▶  $O + O = 10 \times C_1 + R$
- ▶  $C_1 + W + W = 10 \times C_2 + U$
- ▶  $C_2 + T + T = 10 \times C_3 + O$

# Cryptoarithmic Problem

$$\begin{array}{r} \text{ } T \text{ } W \text{ } O \\ + \text{ } T \text{ } W \text{ } O \\ \hline F \text{ } O \text{ } U \text{ } R \end{array}$$



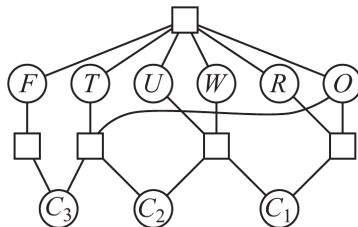
- ▶  $Alldiff(F, T, U, W, R, O)$

It is given that all letters are distinct.

- ▶  $O + O = 10 \times C_1 + R$
- ▶  $C_1 + W + W = 10 \times C_2 + U$
- ▶  $C_2 + T + T = 10 \times C_3 + O$
- ▶  $C_3 = F$

# Cryptoarithmic Problem

$$\begin{array}{r} T \ W \ O \\ + \ T \ W \ O \\ \hline F \ O \ U \ R \end{array}$$



►  $Alldiff(F, T, U, W, R, O)$

It is given that all letters are distinct.

►  $O + O = 10 \times C_1 + R$

►  $C_1 + W + W = 10 \times C_2 + U$

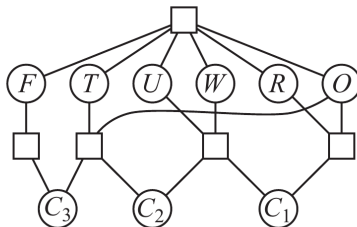
►  $C_2 + T + T = 10 \times C_3 + O$

►  $C_3 = F$  ↗

►  $F \neq 0$  ↗

# Cryptoarithmic Problem

$$\begin{array}{r} T \ W \ O \\ + \ T \ W \ O \\ \hline F \ O \ U \ R \end{array}$$



►  $Alldiff(F, T, U, W, R, O)$

It is given that all letters are distinct.

►  $O + O = 10 \times C_1 + R$

►  $C_1 + W + W = 10 \times C_2 + U$

►  $C_2 + T + T = 10 \times C_3 + O$

►  $C_3 = F$

►  $F \neq 0$

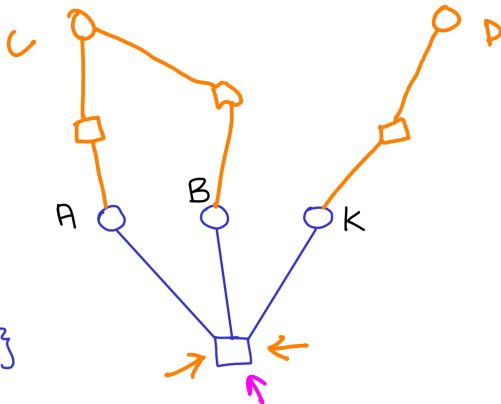
►  $D_i = \{0, \dots, 9\}$

# Constraint satisfaction problem (CSP)

$$X = \{A, B, K\}$$

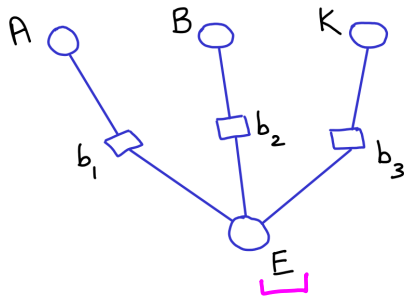
$$D_i = \{1, 2, 3\}$$

$$C = \{A + B = K\}$$

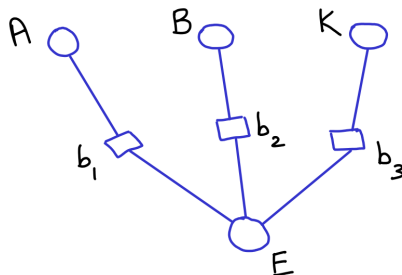




# n-ary constraint to binary constraint



# n-ary constraint to binary constraint

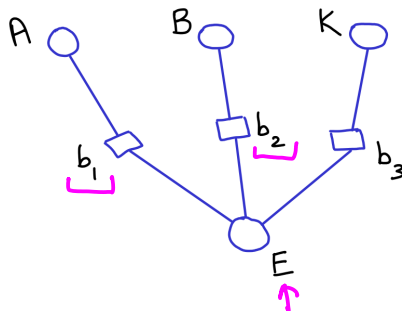


3-ary

►  $D_E = \{ \langle 1, 2, 3 \rangle, \langle 2, 1, 3 \rangle, \langle 1, 1, 2 \rangle \}$

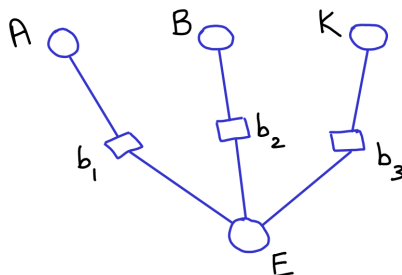
$\underbrace{\hspace{1.5cm}}$        $\uparrow \uparrow \uparrow$        $\uparrow \uparrow \uparrow$        $\uparrow \uparrow \uparrow$

# n-ary constraint to binary constraint



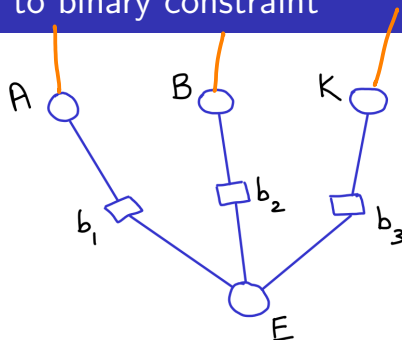
- ▶  $D_E = \{ \langle 1, 2, 3 \rangle, \langle 2, 1, 3 \rangle, \langle 1, 1, 2 \rangle \}$
- ▶ Constraint  $b_1$  checks whether  $A$  equals first element of the tuple assigned to  $E$ .

# n-ary constraint to binary constraint



- ▶  $D_E = \{ \langle 1, 2, 3 \rangle, \langle 2, 1, 3 \rangle, \langle 1, 1, 2 \rangle \}$
- ▶ Constraint  $b_1$  checks whether  $A$  equals first element of the tuple assigned to  $E$ .
- ▶ Constraint  $b_2$  checks whether  $B$  equals second element of the tuple assigned to  $E$ .

# n-ary constraint to binary constraint



- ▶  $D_E = \{ \langle 1, 2, 3 \rangle, \langle 2, 1, 3 \rangle, \langle 1, 1, 2 \rangle \}$
- ▶ Constraint  $b_1$  checks whether A equals first element of the tuple assigned to E.
- ▶ Constraint  $b_2$  checks whether B equals second element of the tuple assigned to E.
- ▶ Constraint  $b_3$  checks whether K equals third element of the tuple assigned to E.

# Dual-graph transformation

- ▶ Add a new variable corresponding to each  $n$ -ary constraint, where  $n > 2$ .

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# Dual-graph transformation

- ▶ Add a new variable corresponding to each  $n$ -ary constraint, where  $n > 2$ .
- ▶ The domain of the new variable consists of  $n$ -tuples whose elements satisfy the corresponding  $n$ -ary constraint.
- ▶ Add a binary constraint between the new variable and each of the original  $n$  variables (that were participating in the original  $n$ -ary constraint).



# Finding a solution for a CSP

- ▶ The problem :  $O(d^n)$  possible assignments.

$$d + d + \dots + d$$

$$\rightarrow d$$

# Finding a solution for a CSP

- ▶ The problem :  $O(d^n)$  possible assignments.
- ▶ Reduce the number of legal values for different variables.

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- ▶ Reduce the number of legal values for different variables.
- ▶ Achieve node consistency ←

$WA \neq \text{Green}$

$D_{WA} = \{r, \cancel{g}, b\}$

# Finding a solution for a CSP

- ▶ The problem :  $O(d^n)$  possible assignments.
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- ▶ Achieve node consistency ←

$WA \neq \text{Green}$

- ▶ Achieve arc consistency ←

