

KRR8: Diagnosis and the rest...

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Outline

- 1 Theory of Reiter
- 2 Diagnosis computation
- 3 Incremental Diagnosis
- 4 And the rest

Reiter Diagnosis

Definition

A **Reiter Diagnosis** for an observed system $(SD, COMP, OBS)$ is a minimal set $\Delta \subset COMP$ such that:

$$SD, OBS, \{\neg Ab(c), c \in COMP \setminus \Delta\}, \{Ab(c), c \in \Delta\}$$

is satisfiable.

Theorem

A **Reiter Diagnosis** is equivalent to a **Minimal Diagnosis**.

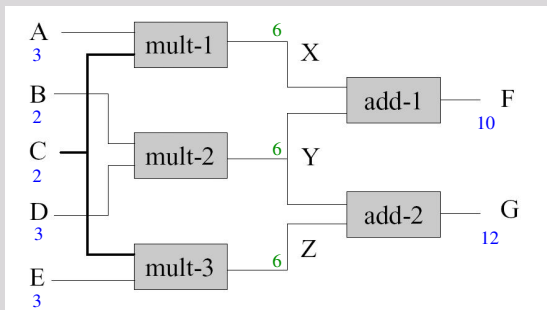
Another representation

An R-diagnosis is seen as a set of components and not a logical sentence. The representation are equivalent.

Reiter Diagnosis: example

Example

Davis circuit



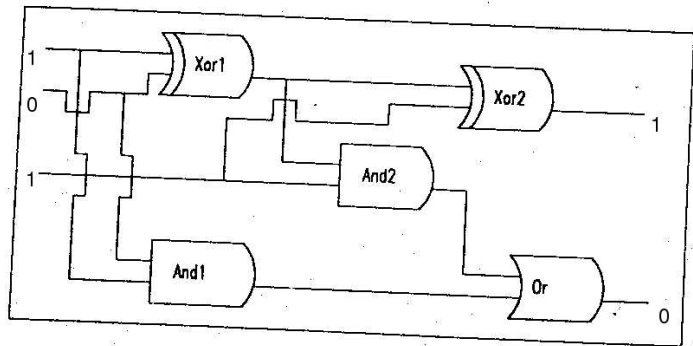
If $OBS = In1(m1, 3), In2(m1, 2), \dots, Out(a2, 12)$ there are 4 R-diagnoses,

$$\{m1\}; \{a1\}; \{m2, m3\}; \{m2, a2\}$$

The R-diagnosis $\{m1\}$ is equivalent to the minimal diagnosis $\neg Ab(a1) \wedge \neg Ab(a2) \wedge Ab(m1) \wedge \neg Ab(m2) \wedge \neg Ab(m3)$.

A new example: an adder

Example



A new example: an adder

Example

SD (behavioural model):

- $AND(x) \wedge \neg Ab(x) \Rightarrow Out(x) = and(In1(x), In2(x))$
- $OR(x) \wedge \neg Ab(x) \Rightarrow Out(x) = or(In1(x), In2(x))$
- $XOR(x) \wedge \neg Ab(x) \Rightarrow Out(x) = xor(In1(x), In2(x))$
- $AND(A1); AND(A2), OR(O1), XOR(X1); XOR(X2)$

SD (structural model):

- $Out(X1) = In2(A2) \dots$

Observations:

- $In1(X1) = 1; In2(X1) = 0; In1(A2) = 1; Out(X2) = 1; Out(O1) = 0.$

R-diagnoses:

- $\{X1\}; \{X2, O1\}; \{X2, A2\}$

Properties of R-Diagnoses

Theorem

\emptyset is the only R-diagnosis for $(SD, COMP, OBS)$ iff

$$SD, OBS, \{\neg Ab(c), c \in COMP\}$$

is satisfiable.

Theorem

$\Delta \subseteq COMP$ is a R-diagnosis iff it is a minimal set such that:

$$SD, OBS, \{\neg Ab(c), c \in COMP \setminus \Delta\}$$

is satisfiable.

How to compute R-diagnoses?

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Definition

An **R-conflict** C is a set $\{c_1, c_2, \dots, c_k\}$ with $c_i \in COMP$ such that:

$$SD, OBS, \{\neg Ab(c), c_i \in C\}$$

is not satisfiable.

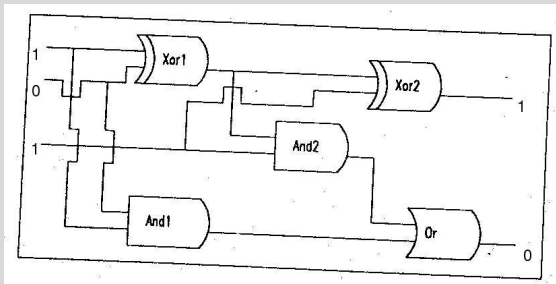
An R-conflict is a set of components $C \subseteq COMP$ which cannot be together in a normal state.

Definition

An **R-conflict** is **minimal** iff there is no strict subset which is also an R-conflict.

R-conflicts: Example 1

Example

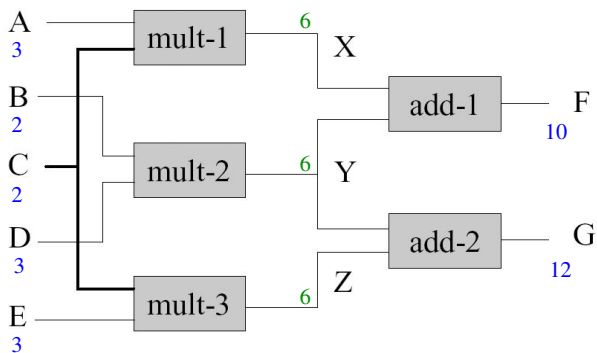


There are 2 minimal R-conflicts:

- 1 ?
- 2 ?

R-conflicts: Example 2

Example



There are 2 minimal R-conflicts:

1 ?

2 ?

R-conflict and R-Diagnosis

Theorem

$\Delta \subseteq \text{COMP}$ is an R-diagnosis for $(\text{SD}, \text{COMP}, \text{OBS})$ iff Δ is a **minimal set** such that $\text{COMP} \setminus \Delta$ is not an R-conflict.

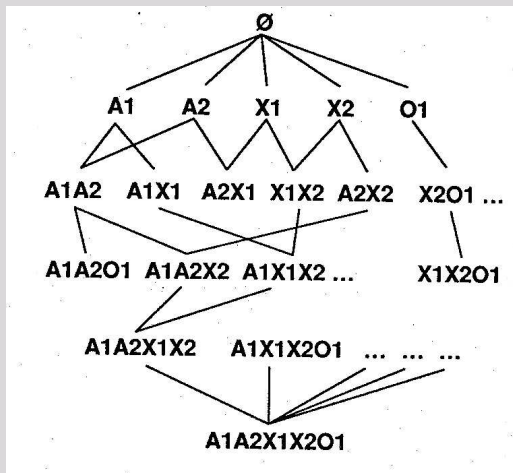
This theorem is the basis of the algorithm DIAGNOSE from Reiter: it is a **lattice** exploration.

Definition

A **lattice** is (roughly) a non-empty partial order set (S, \subseteq) such that every element a, b have an infimum $\text{inf}(a, b)$ (a “lower bound” element) and a supremum $\text{sup}(a, b)$ (an “upper bound” element).

Search space for R-diagnoses

Example



The search space is a **lattice**.

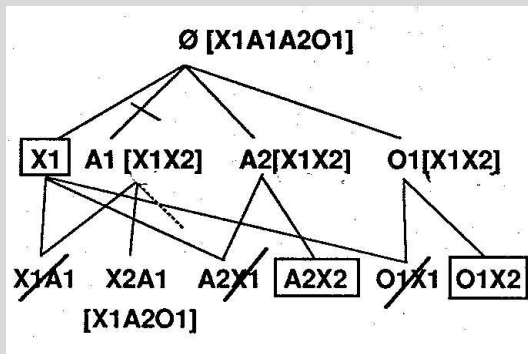
Algorithm

Breadth-first search on the lattice from the empty set \emptyset

- 1 let X the current node in the search
- 2 Call a theorem prover and ask:
Is $COMP \setminus X$ an R-conflict ?
- 3 if **yes**, eliminate the nodes X' such that
 $X' \cap (COMP \setminus X) = \emptyset$
 - X' cannot be a minimal diagnosis.
- 4 if **no**, X is a minimal diagnosis, eliminate the descendants

DIAGNOSE algorithm: example

Example



Sets in brackets are R-conflicts.

Three minimal diagnoses: $\{X1\}$; $\{X2, O1\}$; $\{X2, A2\}$

Another way to solve the problem

The intersection between a diagnosis and any R-conflicts is not empty \Rightarrow **Hitting set**

Theorem

$\Delta \subseteq COMP$ is an **R-diagnosis** for $(SD, COMP, OBS)$ iff Δ is a **minimal hitting set** for the set of minimal conflicts of $(SD, COMP, OBS)$

General diagnosis engine (GDE) from de Kleer.

R-diagnosis: a minimal hitting set problem

Definition

Let $\mathcal{S} = \{S_1, \dots, S_n\}$ be a set of sets, H is a **hitting set** of \mathcal{S} iff

$$H \subseteq_{S_i \in \mathcal{S}} S_i$$

and

$$\forall S_i \in \mathcal{S}, H \cap S_i \neq \emptyset$$

Example

$\mathcal{S} = \{\{a, b\}, \{c, b\}, \{e, f\}\}$ The following sets are hitting sets of \mathcal{S} :

- $H = \{a, b, c, e\}$
- $H = \{b, e\}$ (H is minimal)
- $H = \{a, c, f\}$ (H is minimal)

The following sets are not hitting sets of \mathcal{S} :

- $H = \{a, b\}$
- $H = \{b, e, g\}$

Algorithm

- 1 Computation of all the minimal R-conflicts.
 - Use of an ATMS (Assumption Truth Maintenance System)
 - Update of beliefs about assumptions by retraction of knowledge and declaration of new ones
- 2 Computation of the minimal hitting set on the obtained R-conflicts

R-conflict and R-Diagnosis: examples

Example

Additioner:

The 2 minimal R-conflicts

$\{X1, X2\}$ and $\{X1, A2, O1\}$

correspond to the 3 minimal diagnoses:

$\{X1\}$; $\{X2, O1\}$; $\{X2, A2\}$

Davis circuit:

The 2 minimal R-conflicts:

$\{a1, m1, m2\}$ and $\{a1, a2, m1, m3\}$

correspond to the 4 minimal diagnoses:

$\{m1\}$; $\{a1\}$; $\{a2, m2\}$; $\{m2, m3\}$

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Incremental diagnosis

GDE or DIAGNOSE solve the diagnosis problem in a **off-line** way.

- The observation set is supposed to be **complete**

Observations/Tests

In some systems, an observation is the result of a **test**, an **action**, a **measurement** from the environment to the system.

Definition

The **incremental diagnosis** problem is to:

- compute a diagnosis based on a partial set of observations
- choose what could be the next measurement to perform in the system: **prediction**

Definition

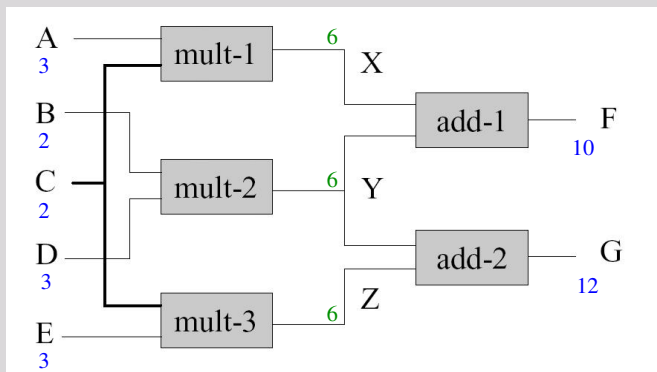
An R-diagnosis Δ predicts O iff

$$SD, OBS, \{\neg Ab(c), c \in COMP \setminus \Delta\}, \{Ab(c), c \in \Delta\} \models O$$

Given the system SD , the current set of observations OBS and the current diagnosis Δ , the system **should produce** the observation O .

Predicted observations: example

Example



- $\Delta_1 : \{m1\}$ predicts $Out(m2) = 6$
- $\Delta_2 : \{m2, m3\}$ predicts $Out(m2) = 4$ and $Out(m3) = 6$.

Updating an R-Diagnosis

Theorem

Confirmation: A R-diagnosis for $(SD, COMP, OBS)$ which predicts O is a R-diagnosis for $(SD, COMP, OBS \wedge O)$.

If the predicted observation O is real (the measurement gives O), then the diagnosis is **confirmed** by the observation O .

Theorem

Invalidation: A R-diagnosis for $(SD, COMP, OBS)$ which predicts $\neg O$ is not a R-diagnosis for $(SD, COMP, OBS \wedge O)$.

If a diagnosis predicts something which is not true, it means that the diagnosis becomes a **wrong hypothesis** and is **invalid**.

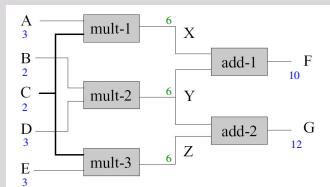
Updating an R-Diagnosis

Algorithm

- 1 Input: $(SD, COMP, OBS)$ an observed system, O a new observation
- 2 Check if Δ predicts O
- 3 if **yes** then Δ is confirmed
 - Δ is a diagnosis of $(SD, COMP, OBS \wedge O)$
- 4 Check if Δ predicts $\neg O$
- 5 If **yes** then look at supersets of Δ

Updating an R-Diagnosis: example

Example



- $\Delta_1 = \{m1\}$ predicts $Out(m2) = 6$
- $\Delta_2 = \{m2, m3\}$ predicts $Out(m2) = 4$
- $\Delta_3 = \{a1\}$ predicts $Out(m2) = 6$
- $\Delta_4 = \{a2, m2\}$ predicts $Out(m2) = 4$

If O is $Out(m2) = 5$, every diagnosis is invalidated. The new ones are supersets: $\{a2, m1, m2\}$, $\{a1, m2, m3\}$, $\{a1, a2, m2\}$, $\{m1, m2, m3\}$

Discriminability/ Diagnosability

Definition

Let O be an observation which confirms Δ_1 and invalidates Δ_2 , we say that O **discriminates**.

Definition

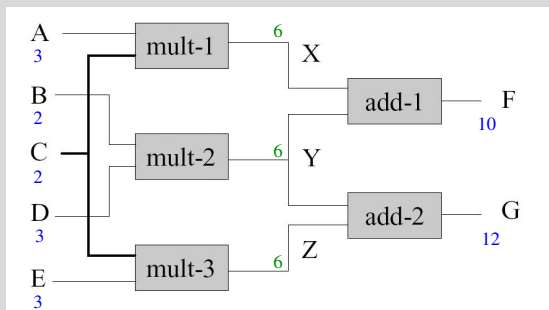
A system $(SD, COMP)$ is **diagnosable** if for any set of possible measurements (any complete set of observations) we have a **unique** diagnosis.

In a diagnosable system, we have enough information (observations) to discriminate between all the diagnoses and to get only one.

Using an incremental diagnosis algorithm on a diagnosable system, we have the guarantee that it **converges** to one diagnosis.

Diagnosability: example

Example



If we can observe only A, B, C, D, E, F, G then the system is **not diagnosable**. If we observe $A, B, C, D, E, F, G, X, Y, Z$ then the system is **diagnosable**.

The observations from B, C, D, E, G do not allow to **discriminate** between diagnoses involving m_2, m_3, a_2 .

- Theory of Reiter: notions of **R-Diagnosis**, **R-conflicts**
- Logic representation \equiv set representations (minimal diagnoses)
- Algorithms:
 - DIAGNOSE: use of a theorem prover, exploration a **lattice**
 - GDE: computation of conflicts and **hitting sets** computation
- **Incremental diagnosis**: update the diagnoses with new measurements
- **Discriminality-Diagnosability** of systems
 - The more information we have, the less numerous are the diagnoses.

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So many things...

- Non-monotonic reasoning
 - **Monotonicity**: if $KB \models \alpha$ then with a new information β , we still have $KB \wedge \beta \models \alpha$
 - The world is full of **exceptions**: every bird can fly, so the emu does!
 - Nonmonotonic logics: **Default logic**, **Circumscription**
- Uncertainty
 - Strong assumption: our knowledge is complete!
 - How to express and make reasoning about **ignorance**, **incompleteness**
 - Use of **probability theory** (Bayesian networks, Markov Decision Process, Fuzzy logic)

So many things...

- Inconsistency
 - Always reasoning with consistency! boring! and bounded! (incompleteness)
 - What about reasoning about inconsistencies: $1 + 1 = 3$ for 1 big enough !
 - **Paraconsistent logics...**
- I give up, I do not have time...