

Computational ‘Consilience’ as a Basis for Theory Formation

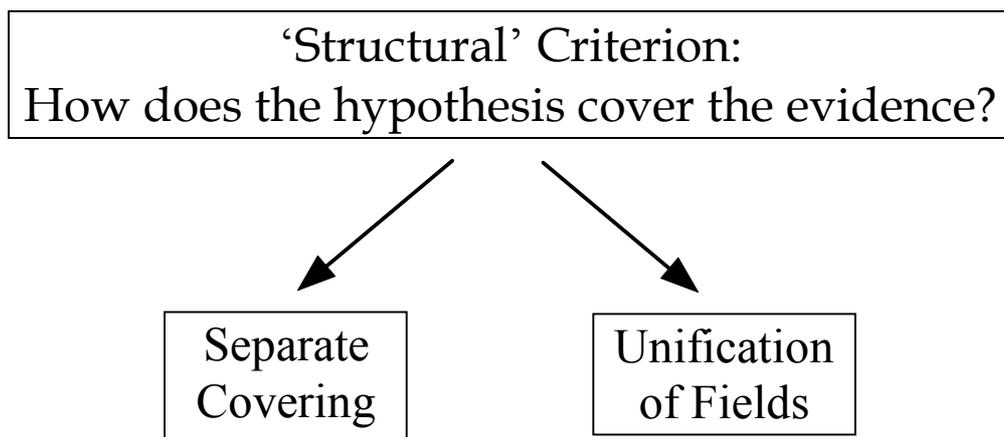
José Hernández-Orallo

Universitat Politècnica de València
Departament de Sistemes Informàtics i Computació
Camí de Vera s/n, E-46022, València, Spain.
E-mail: jorallo@dsic.upv.es.

MODEL-BASED REASONING IN SCIENTIFIC DISCOVERY (MBR'98)
Pavia, Italy, December 17-19, 1998

What is ‘Consilience’?

Term coined in 1847 by Whewell for the selection of inductive theories:



Consilience: the evidence is ‘conciliated’ or unified by the theory.

Related concepts:

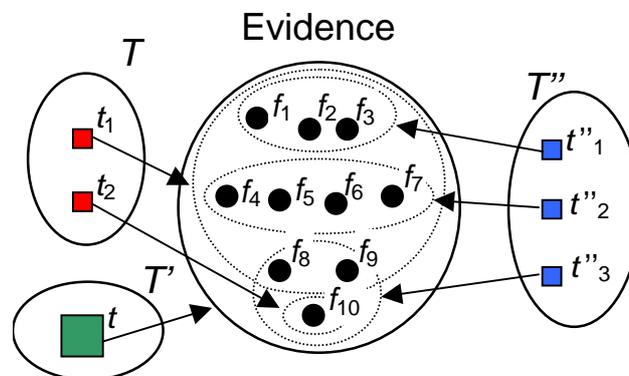
- Principle of ‘Common Cause’ (Reichenbach 1956).
- Explanatory Induction and Scientific Explanation (Harman 1965, Hempel 1965, Ernis 1968).
- Coherence (Thagard 1978).
- Intensionality vs. Tolerance of Partial Extensionality.

Distinguishing Consilience

EXAMPLE:

- Evidence $E = \{ f_1, f_2, \dots, f_{10} \}$
- Hypotheses:

$$T = \{ t_1, t_2 \}, T' = \{ t \} \text{ and } T'' = \{ t''_1, t''_2, t''_3 \}$$



T'' shows two different (but closely related) notions:

- T' and T'' are *intensional*. They have no exceptions.
- T and T'' are *separable*. They are not *consilient*.

Towards Computational Consilience

We denote $M^+(T)$ the model of a theory T .

DEFINITION 1. Separable Theories

A theory T is n -separable in the partition of *different* theories $\Pi = \{ T_1, T_2, \dots, T_n \}$ iff

$$\begin{aligned} M^+(T) &= \bigcup_{i=1..n} M^+(T_i) \\ &\text{and} \\ \forall_{i=1..n} M^+(T_i) &\neq \emptyset \end{aligned}$$

Additional restrictions (*modes*) of separation:

- I. *non-empty*: DEF 1
- II. *non-subset*: DEF 1 + $\forall_{i,j=1..n} (P_i \subseteq P_j \Rightarrow i=j)$.
- III. *disjoint*: DEF 1 + $\forall_{i,j=1..n} (P_i \cap P_j = \emptyset)$.
- IV. *non-subset model*: DEF 1 + $\forall_{i,j=1..n} (M^+(P_i) \subseteq M^+(P_j) \Rightarrow i=j)$.
- V. *disjoint model*: DEF 1 + $\forall_{i,j=1..n} (M^+(P_i) \cap M^+(P_j) = \emptyset)$.

A theory is consilient iff it is not separable.

- The modes give 5 characterisations of consilient theories.

Example

EXAMPLE (Using Horn Theories):

- $P_1 = \{ p(a). q(X) :- r(X). r(a). \}$ is {i-v}separable into $\Pi = \{ \{p(a)\}, \{q(X) :- r(X). r(a)\} \}$.
- $P_2 = \{ q(X) :- r(X). r(b). \}$ is not {ii-v}separable.
- $P_3 = \{ q(X) :- r(X). p(X) :- r(X). r(a). \}$ is non-subset (model) separable into $\Pi = \{ \{ q(X) :- r(X). r(a) \}, \{ p(X) :- r(X). r(a). \} \}$ but it is not disjoint (model) separable.
- $P_4 = \{ q(a). p(X) :- q(X). p(a) \}$ is non-subset (model) and disjoint separable into $\Pi = \{ \{ q(a). p(X) :- q(X). \}, \{ p(a). \} \}$ but it is not disjoint model separable.
But there is $\Pi' = \{ \{ q(a). \}, \{ p(X) :- q(X). p(a). \} \}$
- $P_5 = \{ s(X) :- p(X), q(b). p(X) :- q(X). t(X) :- p(X), q(a) \}$ is non-subset (model) and disjoint separable model into $\Pi = \{ \{ s(X) :- p(X), q(b). p(X) :- q(X) \}, \{ p(X) :- q(X), t(X) :- p(X), q(a) \} \}$ but it is not disjoint separable.

Problems of non-modularity (I, II, IV):

Problems of fantastic consilient concepts (III, IV & V):

P_1 can be 'conciliated' by a fantastic concept f into $P'_1 = \{ p(a) :- f. q(X) :- r(X), f. r(a), f. f. \}$ for iii-iv.

Exception-Free Theories

DEFINITION 2. Exceptions in a Theory

A theory T has $e = \text{card}(M^+(T_E))$ c -exceptions generated from T_E , denoted $\Delta_c(T, T_E) = e$, iff there is a partition $T = \{ T_R, T_E \}$ such that:

$$l(T) - l(T_R) \geq (1 / c) \cdot [l(M^+(T)) - l(M^+(T_R))]$$

where l denotes any syntactical measure of length.

If T_E is not specified, $\Delta_c(T) = \max \{ e \mid \Delta_c(T, T_E) = e \}$

Fixing l and an exception-degree c (usually $c = 1$), a theory T is said to be exception-free iff $\Delta_c(T) = 0$

Pragmatics:

- The modes give 5 characterisations of intensional (exception-free) theories.
- Mode ii and $c=1$ allow modular programs and avoid fantastic concepts.
- For instance, $T = \{ p(X). q(X) \}$ for evidence $\{p(a), p(b), p(e), q(a), q(d), q(e), q(f)\}$ is separable but it has no exceptions.

Reinforcement (1 of 2)

Further detail on the relation hypothesis \leq evidence:

Given a theory, a rule or component r_i is necessary for e iff $T \models e \wedge T - \{ r_i \} \not\models e$

A theory T is reduced for e iff

$T \models e \wedge \neg \exists r_i \in T$ such that it is not necessary for e .

S_1, S_2 are alternative models of T for e iff

$S_1 \subset T, S_2 \subset T, S_1 \neq S_2$ and S_1, S_2 are reduced for e .

We define $Model(e, T)$ as the set of alternative models for example e with respect to T .

We define $Model_r(e, T)$ as the set of alternative models for example e with respect to T that contain r . Formally,

$$Model_r(e, T) = \{ S \subset Model(e, T) \wedge r \in S \}.$$

Reinforcement (2 of 2)

DEFINITION 3. Pure Reinforcement.

The pure reinforcement $\rho(r)$ of a rule r from a theory T wrt. to some given observation $E = \{ e_1, e_2, \dots, e_n \}$ is computed as the number of models of e_i where r is used.

If there are more than one model for a given e_i , *all* of them are reckoned. In the same model, a rule is computed once. Formally,

$$\rho(r) = \sum_{i=1..n} \text{card}(\text{Model}_r(e_i, T))$$

DEFINITION 4. Normalised Reinforcement

$$\rho(r) = 1 - 2^{-\rho(r)}.$$

To avoid fantastic concepts:

DEFINITION 5. Reinforcement wrt. the Data.

The *course* $\chi(f)$ of a given fact f wrt. to a theory is computed as the product of all the reinforcements $\rho(r)$ of all the rules r used in the model of f . If a rule is used more than once, it is computed once. If f has more than one model, we select the greatest course. Formally,

$$\chi(f) = \max_{S \subset \text{Model}(f, T)} \{ \prod_{r \in S} \rho(r) \}$$

Characteristics of Reinforcement

👍 Advantages:

- Reinforcement is easy to compute and allows a flexible evaluation of a theory and the data it covers.
- It provides a measure of the predictive accuracy or assumption feasibility.
- It works for evidences with noise.

👎 Drawbacks:

- For infinite evidence it must be approximated by using a finite sample.

Selection Criteria using Reinforcement

- The Most Reinforced One: The greatest *mean* ($m\chi$) of the courses of all the data presented so far.
- More Compensated: a *geometric mean* instead.

DEFINITION 6. A theory is worthy iff $m\chi(T,E) \geq 0.5$.

If the language is expressible enough there is always a worthy theory for every evidence (just choose every example as an extensional rule).

Easy to define another criteria:

- Intensional: all facts should have a course value greater than the mean divided by a constant (no exceptions).
- Consilience can be better studied: a theory is well-separable if $m\chi$ is not decreased after separation.

Examples:

EXAMPLE 2: (using equational theories)

Consider the following evidence e_1-e_{10} :

$$E = \{ \begin{array}{ll} e_1: e(4) \rightarrow \text{true}, & e_2: e(12) \rightarrow \text{true}, \\ e_3: e(3) \rightarrow \text{false}, & e_4: e(2) \rightarrow \text{true}, \\ e_5: e(7) \rightarrow \text{false}, & e_6: e(7) \rightarrow \text{false}, \\ e_7: e(20) \rightarrow \text{true}, & e_8: e(0) \rightarrow \text{true}, \\ e_9: o(3) \rightarrow \text{true}, & e_{10}: o(2) \rightarrow \text{false} \end{array} \}$$

where natural numbers are represented as e.g. $s(s(s(0)))$ means 3.

$$T_a = \{ \begin{array}{lll} e(s(s(X)) \rightarrow e(X) & : 7 & 0.992 \\ e(0) \rightarrow \text{true} & : 5 & 0.969 \\ e(s(0)) \rightarrow \text{false} & : 3 & 0.875 \\ o(s(s(X)) \rightarrow o(X) & : 2 & 0.75 \\ o(0) \rightarrow \text{false} & : 1 & 0.5 \\ o(s(0)) \rightarrow \text{true} & : 1 & 0.5 \end{array} \}$$

The courses are $\chi(e_1, e_2, e_4, e_7, e_8) = 0.992 \cdot 0.969 = 0.961$, $\chi(e_3, e_5, e_6) = 0.992 \cdot 0.875 = 0.868$, $\chi(e_9) = 0.75 \cdot 0.5 = 0.375$ and $\chi(e_{10}) = 0.75 \cdot 0.5 = 0.375$. **The mean course $m\chi$ is 0.8159.**

$$T_b = \{ \begin{array}{lll} e(s(s(X)) \rightarrow e(X) & : 9 & 0.998 \\ e(0) \rightarrow \text{true} & : 6 & 0.984 \\ e(s(0)) \rightarrow \text{false} & : 4 & 0.938 \\ o(X) \rightarrow \text{not}(e(X)) & : 2 & 0.75 \\ \text{not}(\text{true}) \rightarrow \text{false} & : 1 & 0.5 \\ \text{not}(\text{false}) \rightarrow \text{true} & : 1 & 0.5 \end{array} \}$$

The courses are $\chi(e_1, e_2, e_4, e_7, e_8) = 0.998 \cdot 0.984 = 0.982$, $\chi(e_3, e_5, e_6) = 0.998 \cdot 0.938 = 0.936$, $\chi(e_9) = 0.75 \cdot 0.5 \cdot 0.998 \cdot 0.938 = 0.351$ and $\chi(e_{10}) = 0.75 \cdot 0.5 \cdot 0.998 \cdot 0.984 = 0.368$. **The mean course $m\chi$ is 0.8437.**

Computational Consilience

DEFINITION 7. A theory T is partitionable wrt. an evidence E iff $\exists T_1, T_2 : T_1 \subset T, T_2 \subset T$ and $T_1 \neq T_2$ such that

$$\forall e \in E : T_1 \models e \vee T_2 \models e.$$

We define $E_1 = \{ e \in E : T_1 \models e \}$ and $E_2 = \{ e \in E : T_2 \models e \}$ and $E_{12} = E_1 \cap E_2$.

We will use the term $S\chi(T_1 \oplus T_2, E)$ to denote:

$$m\chi(T_1, E_1) \cdot [\text{card}(E_1) - \text{card}(E_{12})/2] + m\chi(T_2, E_2) \cdot [\text{card}(E_2) - \text{card}(E_{12})/2]$$

DEFINITION 8. A theory T is consilient wrt. an evidence E iff there does not exist a partition T_1, T_2 such that:

$$S\chi(T_1 \oplus T_2, E) \geq m\chi(T, E) \cdot \text{card}(E).$$

In other words, a theory T is consilient wrt. an evidence E iff there does not exist a bipartition $P \in \wp(T)$, such that every example of E is still covered separately without loss of reinforcement.

Computational Consilience (Example)

EXAMPLE 3:

$$E = \{ p(a), p(b), p(e), q(a), q(b), q(e), q(f) \}$$

$$P = \{ p(X) : \rho = 0.875$$

$$q(X) : \rho = 0.9375 \} \quad m\chi(E, P) =$$

0.9107

The partition:

$$P_1 = \{ p(X) : \rho = 0.875 \} \quad m\chi(E_1, P_1) = 0.875$$

$$P_2 = \{ q(X) : \rho = 0.9375 \} \quad m\chi(E_2, P_2) = 0.9375$$

$$S\chi(P_1 \oplus P_2, E) = m\chi(E_1, P_1) \cdot 3 + m\chi(E_2, P_2) \cdot 4 =$$

$$m\chi(E, P) \cdot \text{card}(E) =$$

0.9107 · 7.

P is not consilient.

Definition 8 can be parameterised by introducing a consilience factor.

Intrinsic Exceptions and Consilience

An intrinsic exception or extensional patch is defined as a rule r with $\rho = 0.5$, i.e. a rule that just covers one example e .

We must distinguish between:

- *completely extensional exceptions*, when r does not use any rule from the theory to cover e ,
- *partially extensional exceptions* when r uses other rules to describe e .

Theorem 1. If a worthy theory T for an evidence E has a rule r with $\rho = 0.5$, and completely extensional, then T is not consilient.

This justifies the use of consilience as the motor or maxim of theory formation:

If part of the evidence is covered extensionally, a revision should be made to conciliate it with the rest.

For explanatory induction, not only prediction errors or anomalies motivate theory revision.

Proof of Theorem 1

Just choose the partition $T_1 = T - r$ and $T_2 = T$. Since $\rho = 0.5$ then r is only used by one example e_r . Since it is a completely extensional exception, we have that r does not use any rule from T_1 to cover e_r , so $\rho'(r_i) = \rho(r_i)$ for all $r_i \in T_1$. Let n be the number of the examples of the evidence E . Hence, $m\chi(T_1, E_1) = [m\chi(T, E) \cdot n - \chi(e_r, T)] / (n-1) = [m\chi(T, E) \cdot n - 1/2] / (n-1) = [m\chi(T, E) \cdot n + m\chi(T, E) - m\chi(T, E) - 1/2] / (n-1) = m\chi(T, E) + [m\chi(T, E) - 1/2] / (n-1)$.

From def. 7, the inequality simplifies as follows:

$$\begin{aligned}
 S\chi(T_1 \oplus T_2, E) &= \\
 m\chi(T_1, E_1) \cdot [\text{card}(E_1) - \text{card}(E_{12})/2] &+ m\chi(T_2, E_2) \cdot [\\
 \text{card}(E_2) - \text{card}(E_{12})/2] &= \\
 \{ m\chi(T, E) + [m\chi(T, E) - 1/2] / (n-1) \} \cdot [(n-1) - (n-1)/2] &+ m\chi(T, E) \cdot [n - (n-1)/2] = \\
 m\chi(T, E) \cdot [(n-1) - (n-1)/2 + n - (n-1)/2] &+ [m\chi(T, E) - \\
 1/2] \cdot [(n-1) - (n-1)/2] / (n-1) &= \\
 m\chi(T, E) \cdot [n] + [m\chi(T, E) - 1/2] / 2 &
 \end{aligned}$$

Since T is worthy, then $m\chi(T, E) \geq 0.5$, and finally

$$S\chi(T_1 \oplus T_2, E) \geq m\chi(T, E) \cdot n = m\chi(T, E) \cdot \text{card}(E). \quad \square$$

Consilience and Coherence

Thagard's *modern* view of coherence is equivalent to constraint satisfaction wrt. the background knowledge B .

Coherence allows the analysis of:

- Deductive Compatibility with B .
- Explanatory Compatibility with B . (abduction or nomological induction)

However, it is not constructive, so it is not useful for non-nomological induction → the idea of satisfaction is not clear...

Consilience represents this notion of 'accordance' with the background knowledge B , by measuring the direct or constructive inter-relation with B .
--

Conclusions

Formalising 'Consilience':

- First approach to consilience based on model partition.
- Second approach based Reinforcement: Further detail on the relation between hypothesis and evidence.

From here we have shown that:

- Consilience is different (but related) that intentionality.
- Consilience can be used to detect which parts of the theory are weak.
- Consilience and Coherence are somehow *complementary*.

<p><i>A consilient model is the goal of theory construction. Exceptional or unconsilient parts should trigger</i></p>

Notes:

* Slide "Towards Computational Consilience"

(If T was a Logic Theory, $M^+(T)$ could be the Minimal Herbrand Model of T).