

Revision of Bayesian Networks

Belief Revision in Philosophy

Belief: State of mind, that something is the case - often degrees of beliefs are used

There are several different ways for a **Belief Change:**

Contraction: removal of a belief;

Expansion: addition of a belief without checking consistency;

Revision: addition of a belief while maintaining consistency;

Extraction: extracting a consistent set of beliefs and/or epistemic entrenchment ordering;

Consolidation: restoring consistency of a set of beliefs;

Merging: fusion of two or more sets of beliefs while maintaining consistency.

There are postulates for a **logical revision operator** that a rational operator should satisfy (Gärdenfors 1985). But how to change **Belief degrees**?

Example for a **Revision**: It is known that a certain navigation system can only be included in the car if one of the corresponding radio systems is already installed. It is planned to sell 3000 instead of 1000 navigation systems in the next quartal. How many radio systems of each type should be bought?

Revision of Probabilistic Graphical Models

Graphical models are efficient for representing domain knowledge. After some time additional observations can change our underlying knowledge of the domain.

We need a way to incorporate belief changes to avoid updating the whole knowledge base.

Idea: Local changes should only lead to local adaptations of the knowledge base and necessary consequences.

Revision of Probabilistic Graphical Models

Prior Probability
Distribution

New conditional
probabilities



Revision

necessary changes, but **minimal** changes



Posterior Probability Distribution including specified and inferred changes

Solution: Search for the Information-theoretically closest distribution to the prior distribution that satisfies the new knowledge

Revision Operator - Iterative Proportional Fitting

Iterative proportional fitting (raking, matrix scaling), is a well-known algorithm (1937, often reinvented) for adapting the marginal distributions of a given joint distribution to desired values.

It consists in computing the following sequence of probability distributions:

$$p_U^{(0)}(u) \equiv p_U(u) \quad (1)$$

$$\forall 1, 2, \dots : p_U^{(i)}(u) \equiv p_U^{(i-1)}(u) \frac{p_{A_j}(a)}{p_{A_j}^{(i-1)}(a)} \quad (2)$$

In each step the probability distribution is modified in such a way that the resulting distribution satisfies the given marginal distribution A_j . However, this will, in general, change the marginal distribution for an earlier adapted variable A_k .

Therefore, the adaptation has to be iterated, traversing the set of variables several times. The process is proofed to converge for non-contradicting revision statements, and **only for statements with no inconsistencies.**

Revision Algorithm

The revision algorithm sums up as follows:

```
1: forall  $C \in \mathcal{C}$  do  
2:    $p_C^{(0)}(c) \equiv p_C(c)$   
3:    $i \equiv 0$   
4:   repeat  
5:      $i \equiv i + 1$ ;  
6:     forall  $C \in \mathcal{C}$  do  
7:       forall  $j \in J_C$  do  
8:          $p_C^{(i)}(c) \equiv p_C^{(i-1)}(c) \frac{p_{A_j}(a)}{p_{A_j}^{(i-1)}(a)}$ ;  
9:       do evidence propagation  
10:    end  
11:  until convergence
```

See https://en.wikipedia.org/wiki/Iterative_proportional_fitting for more details

Inconsistencies

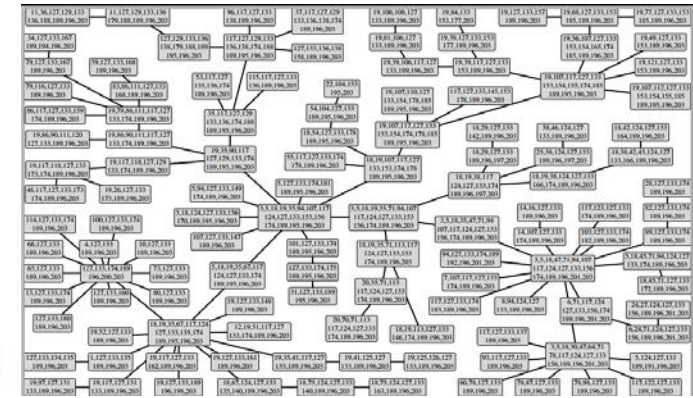
Inconsistencies can emerge in the presence of:

Complex structure: dependencies between attributes

e.g. dependencies between car components

Many revision assignments: changes of the probability distribution

e.g. changing installation rate of component combinations



Inconsistencies are unincorporatable changes / inconsistent revision assignments

	3490	SL	✓	DC	...	100001	HOCH
	3490	SL	✓	DC	...	100002	HOCH
(Bereich:EXPORT)	2729	SL	✓	DC	...	100003	HOCH
(Bereich:TR AND)	721	SL	✓	DC	...	100004	HOCH
(Bereich:EXPORT) (&.and;ARGENTIN...	18	SL	✓	DC	...	100005	HOCH
(Bereich:EXPORT) (&.and;AUSTRAL...	43	SL	✓	DC	...	100006	HOCH
(Bereich:EXPORT) (&.and;SALTIMB)	2	SL	✓	DC	...	100007	HOCH
(Bereich:EXPORT) (&.and;BELGIEN)	180	SL	✓	DC	...	100008	HOCH
(Bereich:EXPORT) (&.and;BRASILIEN)	15	SL	✓	DC	...	100009	HOCH
(Bereich:EXPORT) (&.and;BULGARI...	2	SL	✓	DC	...	100010	HOCH
(Bereich:EXPORT) (&.and;CHINA)	0	SL	✓	DC	...	100011	HOCH
(Bereich:EXPORT) (&.and;DÄNERF...	47	SL	✓	DC	...	100012	HOCH
(Bereich:EXPORT) (&.and;EXPRESS)	0	SL	✓	DC	...	100013	HOCH
(Bereich:EXPORT) (&.and;EXPRESS...	0	SL	✓	DC	...	100014	HOCH
(Bereich:EXPORT) (&.and;EXPRESS...	0	SL	✓	DC	...	100015	HOCH
(Bereich:EXPORT) (&.and;EXPRESS...	0	SL	✓	DC	...	100016	HOCH
(Bereich:EXPORT) (&.and;EXPRESS...	0	SL	✓	DC	...	100017	HOCH
(Bereich:EXPORT) (&.and;EXPRESS...	0	SL	✓	DC	...	100018	HOCH
(Bereich:EXPORT) (&.and;FINLAND)	26	SL	✓	DC	...	100019	HOCH
(Bereich:EXPORT) (&.and;FRANKRE...	230	SL	✓	DC	...	100020	HOCH
(Bereich:EXPORT) (&.and;GRIECH...	7	SL	✓	DC	...	100021	HOCH
(Bereich:EXPORT) (&.and;GROSSBR...	978	SL	✓	DC	...	100022	HOCH
(Bereich:EXPORT) (&.and;HONGK...	8	SL	✓	DC	...	100023	HOCH
(Bereich:EXPORT) (&.and;INDIEN)	0	SL	✓	DC	...	100024	HOCH
(Bereich:EXPORT) (&.and;IRLAND)	7	SL	✓	DC	...	100025	HOCH
(Bereich:EXPORT) (&.and;ISLAND)	0	SL	✓	DC	...	100026	HOCH
(Bereich:EXPORT) (&.and;ISRAEL)	6	SL	✓	DC	...	100027	HOCH
(Bereich:EXPORT) (&.and;ITALIEN)	265	SL	✓	DC	...	100028	HOCH
(Bereich:EXPORT) (&.and;JAPAN)	96	SL	✓	DC	...	100029	HOCH
(Bereich:EXPORT) (&.and;KANADA)	0	SL	✓	DC	...	100030	HOCH
(Bereich:EXPORT) (&.and;KANAREN)	2	SL	✓	DC	...	100031	HOCH
(Bereich:EXPORT) (&.and;KATATH)	3	SL	✓	DC	...	100032	HOCH
(Bereich:EXPORT) (&.and;LUXEMB...	12	SL	✓	DC	...	100033	HOCH
(Bereich:EXPORT) (&.and;MALAYSI)	0	SL	✓	DC	...	100034	HOCH
(Bereich:EXPORT) (&.and;MEXIKO)	0	SL	✓	DC	...	100035	HOCH
(Bereich:EXPORT) (&.and;NIELSREL...	7	SL	✓	DC	...	100036	HOCH
(Bereich:EXPORT) (&.and;NIEDER...	95	SL	✓	DC	...	100037	HOCH
(Bereich:EXPORT) (&.and;NORWEG...	39	SL	✓	DC	...	100038	HOCH
(Bereich:EXPORT) (&.and;O) AFRIKA)	3	SL	✓	DC	...	100039	HOCH
(Bereich:EXPORT) (&.and;O) EURO...	1	SL	✓	DC	...	100040	HOCH
(Bereich:EXPORT) (&.and;OESTERR...	65	SL	✓	DC	...	100041	HOCH
(Bereich:EXPORT) (&.and;POLLEN)	10	SL	✓	DC	...	100042	HOCH
(Bereich:EXPORT) (&.and;PORTUGAL)	24	SL	✓	DC	...	100043	HOCH
(Bereich:EXPORT) (&.and;R) AGCC)	0	SL	✓	DC	...	100044	HOCH
(Bereich:EXPORT) (&.and;R) (AS) PAZ)	1	SL	✓	DC	...	100045	HOCH
(Bereich:EXPORT) (&.and;R) ASEAN)	0	SL	✓	DC	...	100046	HOCH
(Bereich:EXPORT) (&.and;R) (NAHOST)	0	SL	✓	DC	...	100047	HOCH
(Bereich:EXPORT) (&.and;R) (SUDAN)	12	SL	✓	DC	...	100048	HOCH
(Bereich:EXPORT) (&.and;RUMÄN...	3	SL	✓	DC	...	100049	HOCH
(Bereich:EXPORT) (&.and;RUSSLAND)	61	SL	✓	DC	...	100050	HOCH
(Bereich:EXPORT) (&.and;S) AFRICA)	42	SL	✓	DC	...	100051	HOCH
(Bereich:EXPORT) (&.and;S) SAUDI A...	0	SL	✓	DC	...	100052	HOCH
(Bereich:EXPORT) (&.and;SCHWED...	69	SL	✓	DC	...	100053	HOCH
(Bereich:EXPORT) (&.and;SCHWEIZ)	64	SL	✓	DC	...	100054	HOCH

Inner Inconsistencies

Revision assignments are inconsistent independent of prior distribution

Outer Inconsistencies

Revision assignments inconsistent with zero-values in prior distribution

Example: Inner Inconsistencies

Insert revision assignments in probability distribution:

0.6	0.3	
0.2	0.25	
0.3		
		0.1

0.5

Inner inconsistencies can emerge as consequences of probability implications:

0.6	0.3	0.1	$\leftarrow 0.10 = 1 - 0.6 - 0.3$
0.2	0.25		
0.3			
0.1		0.1	$\leftarrow 0.10 = 0.6 - 0.2 - 0.3$

0.5

Example: Inner Inconsistencies

Insert revision assignments in probability distribution:

0.6	0.3		
0.2	0.25		0.5
0.3			
		0.1	

Inner inconsistencies can emerge as consequences of probability implications:

0.6	0.3	0.1	
0.2	0.25	0.0	0.5
0.3		0.0	
0.1		0.1	

← set to zero since column sum is already maximum
← set to zero since column sum is already maximum

Example: Inner Inconsistencies

Insert revision assignments in probability distribution:

0.6	0.3		
0.2	0.25		0.5
0.3			
		0.1	

Inner inconsistencies can emerge as consequences of probability implications:

0.6	0.3	0.1	
0.2	0.25	0.0	0.45 ← $0.45 = 0.2 + 0.25 + 0.0$
0.3		0.0	0.5
0.1		0.1	0.05

Example: Inner Inconsistencies

Insert revision assignments in probability distribution:

0.6	0.3		
0.2	0.25		0.5
0.3			
		0.1	

Inner inconsistencies can emerge as consequences of probability implications:

0.6	0.3	0.1	
0.2	0.25	0.0	0.45
0.3		0.0	0.5
0.1		0.1	?

Contradicting implications: $0.05 \neq 0.20$

column-sum $\Rightarrow 1 - 0.45 - 0.5 = 0.05$

row-sum $\Rightarrow 0.1 + 0.1 = 0.20$

Example: Outer Inconsistencies

Insert revision assignments in probability distribution with fixed zero values (\times):

0.1	0.5	0.4	
	\times	\times	0.3
	\times		0.5

Outer inconsistencies can emerge as consequences of probability implications:

0.1	0.5	0.4	
	\times	\times	0.3
	\times		0.5
	0.5		0.2

Contradicting implications: $0.5 > 0.2$

$$\text{column-sum} \Rightarrow 0.5 - 0.0 - 0.0 = 0.5$$

$$\text{column-sum} \Rightarrow 1.0 - 0.3 - 0.5 = 0.2$$

Systematic Handling of Revision Inconsistencies

Even for an expert user it is not easy to configure revision statements without creating inconsistencies!

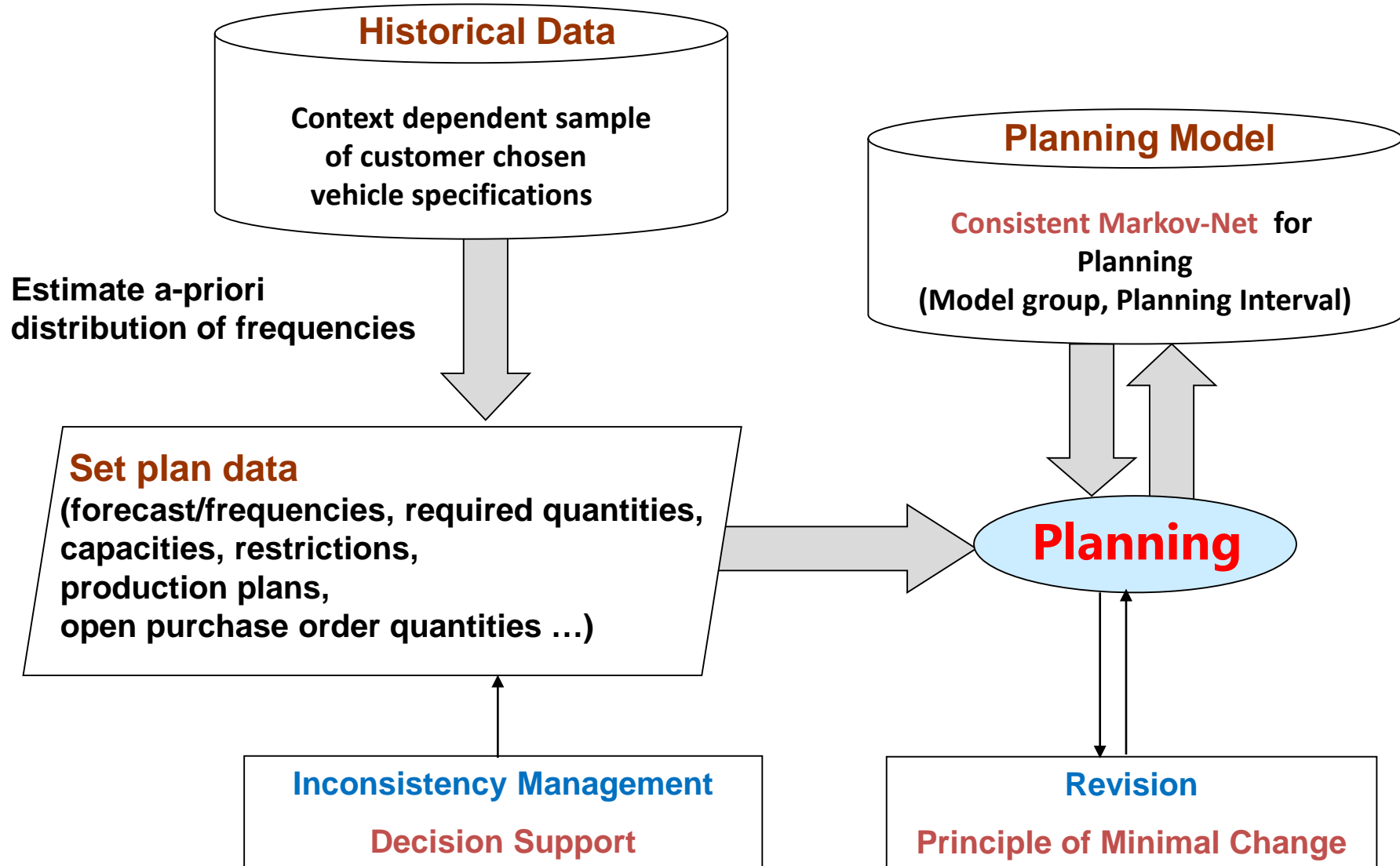
In the case of belief without uncertainty there are methods to handle inconsistencies: The **Logics** of Formal **Inconsistency** (LFIs) are a family of paraconsistent **logics** that constitute consistent fragments of classical **logic** yet which reject the explosion principle where a **contradiction** is present.

The case of graded belief is more complicated:

If the Revision-Operation fails, we need to explain the user how to change his desired revision statements. Otherwise no solution can be found.

There must be an inconsistency management before using the revision operator. In our real example it looks as follows:

Example: Planning Operation Revision



Systematic Handling of Revision Inconsistencies

Example for one planning week

Data volume (cumulative per week)	Across all 166 Model groups	Single Model group
Number of Markov Nets	75.787	1.054

... for one planning interval	Model group 1	Model group 2
Planning requirement	4.424	1.299
Number of variables	203	204
Number of cliques	174	156

Largest clique (only positive probabilities)	130.806 tupels 9 variables	1.489.515 tupels 14 variables
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