Decision Support via Expert Systems

6.872/HST950

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Components of an Expert System

•Knowledge

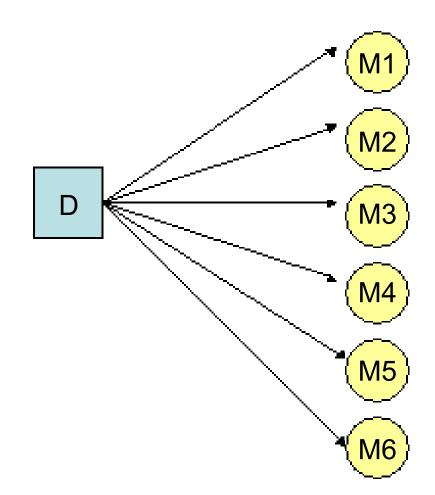
-In various forms: associations, models, etc.

- •Strategy –Baconian, exhaustive enumeration, on-line, etc.
- •Implementation –Programs, pattern matching, rules, etc.

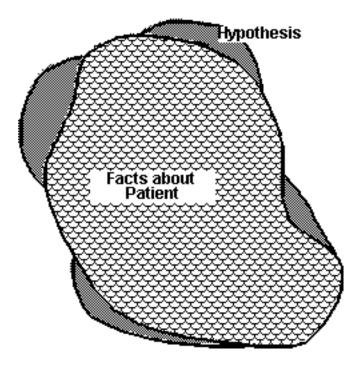
Last Time

Naïve Bayesian Inference

- Exhaustive and Mutually
 Exclusive disease
 hypotheses (1 and only 1)
- Conditionally independent observables (manifestations)
- $-P(D_i), P(M_{ij}|D_i)$



Taking the Present Illness—Diagnosis by Pattern Directed Matching



PIP's Theory of Diagnosis •From initial complaints, *guess* suitable hypothesis.

•Use current active hypotheses to guide questioning

•Failure to satisfy expectations is the strongest clue to a better hypothesis; *differential diagnosis*

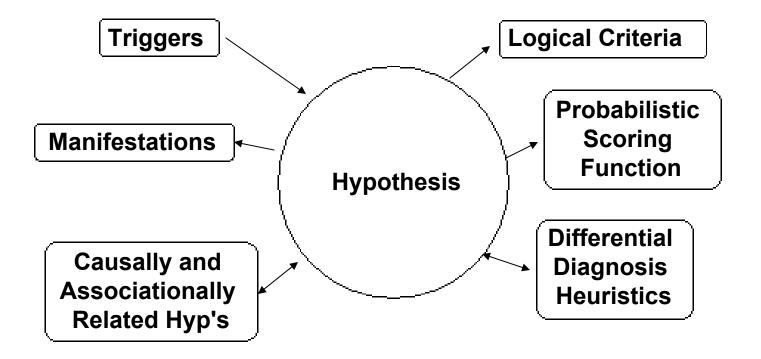
Hypotheses are activated, de-activated, confirmed or rejected based on

 -(1) logical criteria
 -(2) probabilities based on:

findings local to hypothesis

causal relations to other hypotheses

Memory Structure in PIP



PIP's Model of Nephrotic Syndrome

•NEPHROTIC SYNDROME, a clinical state •FINDINGS:

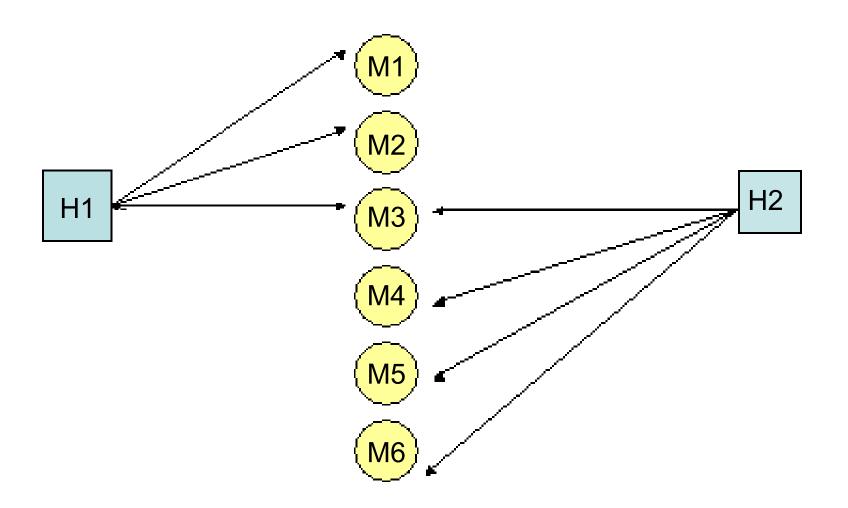
- -1* Low serum albumin concentration
- -2. Heavy proteinuria
- -3* >5 gm/day proteinuria
- -4* Massive symmetrical edema
- -5* Facial or peri-orbital symmetric edema
- -6. High serum cholesterol
- -7. Urine lipids present
- *IS-SUFFICIENT:* Massive pedal edema & >5 gm/day proteinuria
 MUST-NOT-HAVE: Proteinuria absent

•SCORING . . .

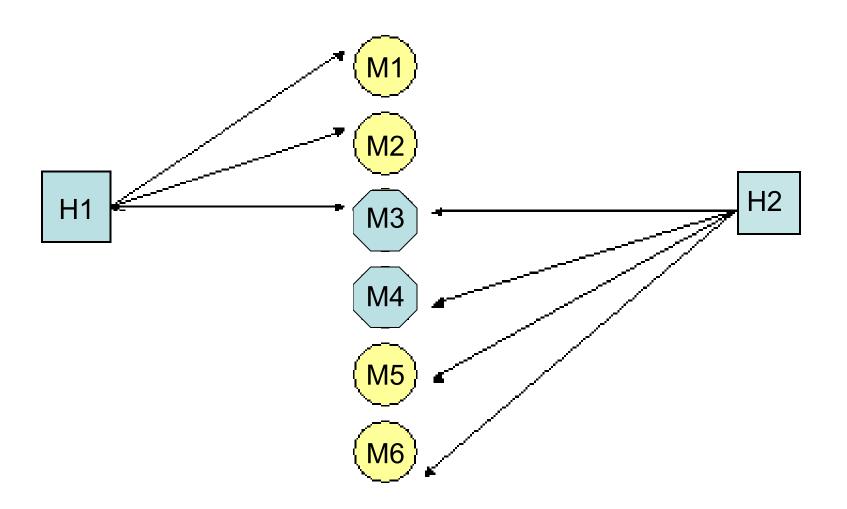
 •MAY-BE-CAUSED-BY: AGN, CGN, nephrotoxic drugs, insect bite, idiopathic nephrotic syndrome, lupus, diabetes mellitus
 •MAY-BE-COMPLICATED-BY: hypovolemia, cellulitis
 •MAY-BE-CAUSE-OF: sodium retention
 •DIFFERENTIAL DIAGNOSIS:

-neck veins elevated

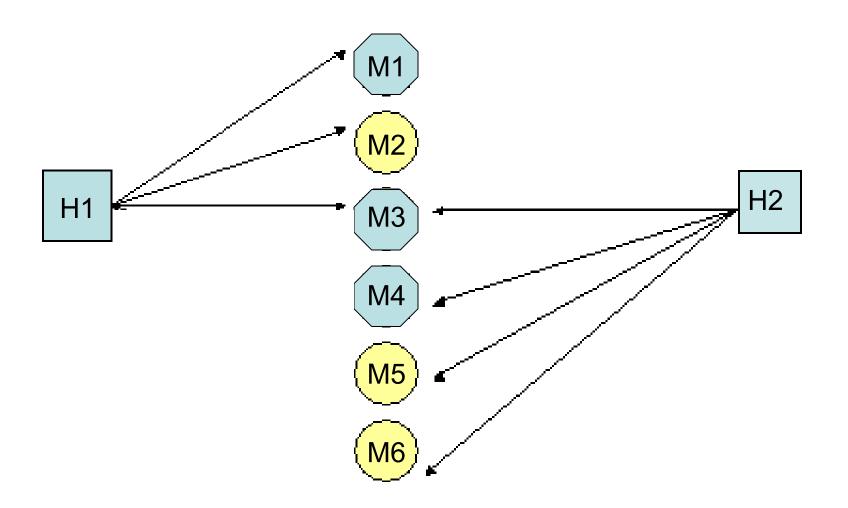
QMR Partitioning



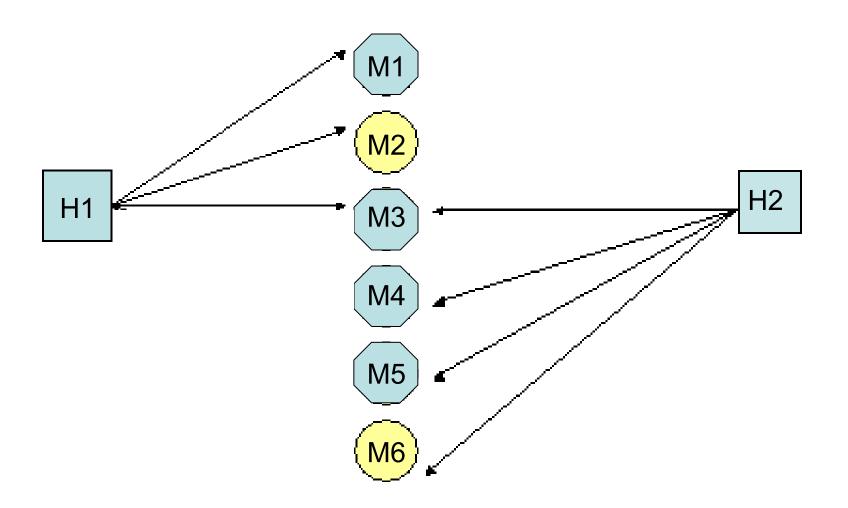
Competitors



Still Competitors



Still Competitors



Multi-Hypothesis Diagnosis

- •Set aside complementary hypotheses
- •... and manifestations predicted by them
- •Solve diagnostic problem among competitors
- •Eliminate confirmed hypotheses and manifestations explained by them
- •Repeat as long as there are coherent problems among the remaining data

Internist/QMR

•Knowledge Base:

- •956 hypotheses
- •4090 manifestations (about 75/hypothesis)
- Evocation like P(H|M)
- •*Frequency* like P(M|H)
- •Importance of each M
- •Causal relations between H's
- •Diagnostic Strategy:
 - Scoring function
 - •Partitioning
 - •Several questioning strategies

QMR Scoring

Positive Factors

- •Evoking strength of observed Manifestations
- Scaled Frequency of causal links from
- confirmed Hypotheses

Negative Factors

- •Frequency of predicted but absent
- Manifestations
- Importance of unexplained Manifestations

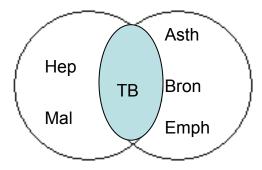
•Various scaling parameters (roughly exponential)

Symptom Clustering for Multi-Disorder Diagnosis

- Tom Wu, Ph.D. 1991

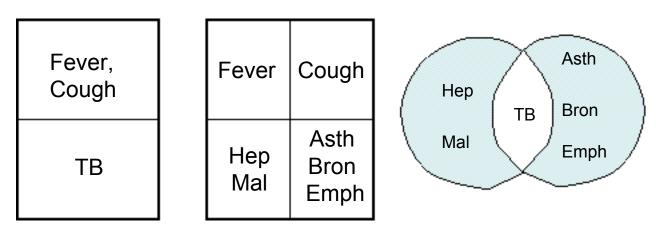
Clustering Alternatives

Symptom	Possible Causes
Fever	TB, Hepatitis, Malaria
Cough	TB, Asthma, Bronchitis, Emphysema



H1

H2



Symptom Clustering is Efficient

Like in any "planning island" approach, reducing an exponential problem to several smaller exponential problems vastly improves efficiency, *if it captures some insight into the problem.*

Wu's algorithm (SYNOPSIS) will keep a compact encoding even if it overgenerates slightly.

E.g., suppose that of the set of diseases represented by (d5, d6) x (d3, d7, d8, d9) x (d1, d2, d4), d6 x d8 x d1 is not a candidate. To represent this precisely would require enumerating the 23 valid candidates. Instead, the factored representation is kept.

In a diagnostic problem drawn from a small subset of the Internist database, it is a *power of 3* faster and a *power of 5* more compact than standard symptom clustering.

Guide search via probabilities, if we have a reasonable model(!)