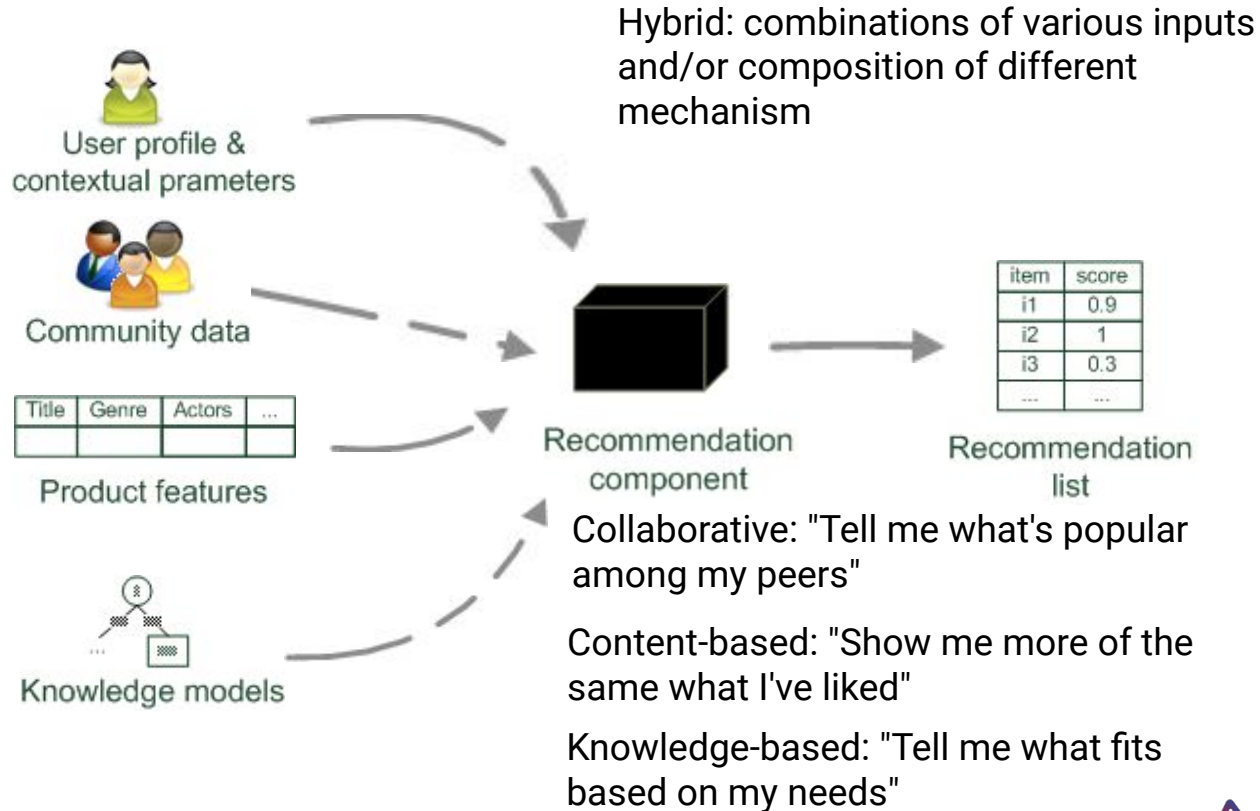


Hybrid recommender systems

Hybrid recommender systems

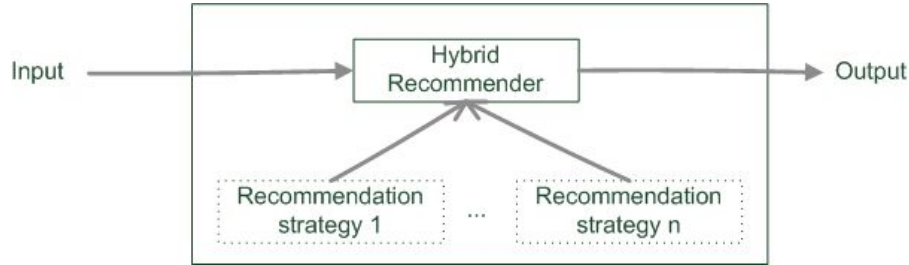


Hybrid recommender systems

- All three base techniques are naturally incorporated by a good sales assistant but have their shortcomings
 - For instance, cold start problems
- Idea of crossing two (or more) species/implementations
 - *hybrida* [lat.]: denotes an object made by combining two different elements
 - Avoid some of the shortcomings
 - Reach desirable properties not (or only inconsistently) present in parent individuals

Monolithic hybridization design

- Only a single recommendation component



- Features/knowledge sources of different paradigms are combined

Monolithic hybridization design

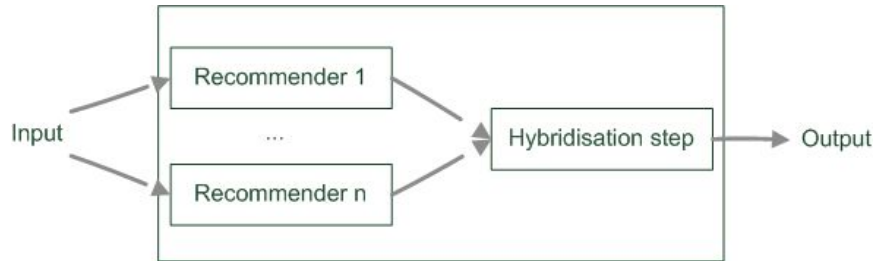
User	Item1	Item2	Item3	Item4	Item5	Item	Genre
Alice		1		1		Item1	romance
User1		1	1		1	Item2	mystery
User2	1	1			1	Item3	mystery
User3	1		1			Item4	mystery
User4					1	Item5	fiction



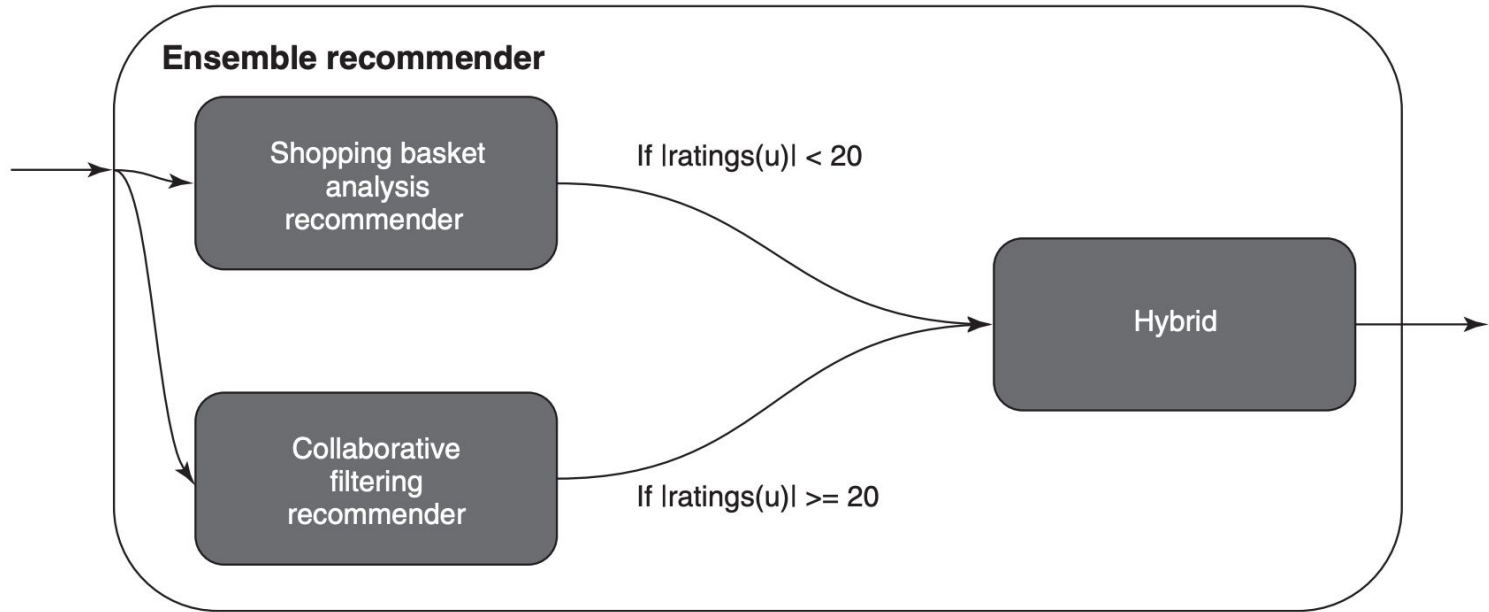
Feature	Alice	User1	User2	User3	User4
User likes many <i>mystery</i> books	true	true			
User likes some <i>mystery</i> books			true	true	
User likes many <i>romance</i> books					
User likes some <i>romance</i> books			true	true	
User likes many <i>fiction</i> books					
User likes some <i>fiction</i> books		true	true		true

Parallelized hybridization design

- Output of several existing implementations combined
- Some weighting or voting scheme
 - Weights can be assigned to each method
 - Switching in which based on a condition we switch to a different recommender



Parallelized hybridization design: Switched Ensemble



Parallelized hybridization design: Weighted

- Compute weighted sum: $rec_{weighted}(u,i) = \sum_{k=1}^n \beta_k \times rec_k(u,i)$

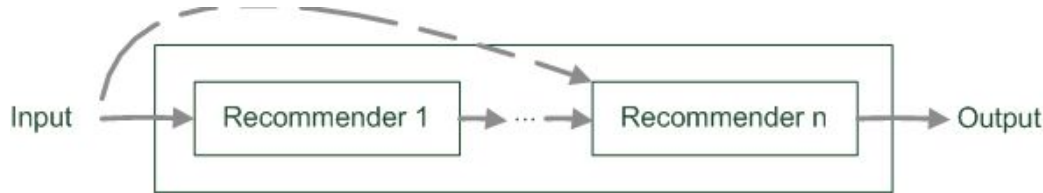
Recommender 1		
Item1	0.5	1
Item2	0	
Item3	0.3	2
Item4	0.1	3
Item5	0	

Recommender 2		
Item1	0.8	2
Item2	0.9	1
Item3	0.4	3
Item4	0	
Item5	0	

Recommender weighted(0.5:0.5)		
Item1	0.65	1
Item2	0.45	2
Item3	0.35	3
Item4	0.05	4
Item5	0.00	

Pipelined hybridization designs

- One recommender system pre-processes some input for the subsequent one
- Refinement of recommendation lists (cascade)
- One recommender builds a model that is exploited by the principal recommender to make recommendations. (Meta-level)



What's Next?