Automated Narrative Planning Model Extension

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What is Interactive Narrative?

- Automated storytelling where characters' actions are generated automatically
- Uses AI planning instead of branching story paths
- Challenges:
  - Must respond to user choices (robustness)
  - Must tell new, interesting stories on replay (diversity)
- Currently, humans must manually write all narrative actions which is too time consuming

The Core Idea

- Problem: domain solution is a bottleneck
- Solution: Automatically expand a small, human-made narrative using:
  - Anton - generates opposite actions
  - Thype - generates new types based on hierarchy

How does “anton” work?

- Analyzes existing actions to detect missing *reverse* transitions
- Uses linguistic resources to find antonyms
- Generates human-readable new actions
- Example:
  - love-spell
  - Anton generates: hate-spell
- Enables story recovery when the user disrupts the plan

How does “thype” work?

- Looks at type hierarchies (like “person → king → emperor”)
- Uses WordNet and ConceptNet to generate related types (hyponyms/hypernyms)
- Adds new types with minimal human input
- Enables variety by introducing new story elements.

Type	Action	State Transition Rules: $E \Rightarrow S \rightarrow F$					Rule #
person	love-spell	$alive_1$	$\Rightarrow$	$\neg loves_1$	$\rightarrow$	$loves_1$	1
		$alive_2$	$\Rightarrow$	$\neg loves_2$	$\rightarrow$	$loves_2$	2
	summon	$has_2\ at_1$	$\Rightarrow$	$\{\}$	$\rightarrow$	$controls_1$	3
monster	slay	at_1	$\Rightarrow$	$alive_1$	$\rightarrow$	$\{\}$	4
male	marry	$loves_1$, $loves_2$, ..	$\Rightarrow$	$single_1$	$\rightarrow$	$married_1$ $married_2$	5
	fall-in-love	$alive_1$	$\Rightarrow$	$\neg loves_1$	$\rightarrow$	$loves_1$	6
princess	marry	$loves_2$ $loves_1$...	$\Rightarrow$	$single_1$	$\rightarrow$	$married_2$ $married_1$	7
	fall-in-love	$beautiful_1$ at_1 $alive_1$	$\Rightarrow$	$\neg loves_2$	$\rightarrow$	$loves_2$	8
genie	summon	$alive_1$	$\Rightarrow$	in_1 $confined_1$	$\rightarrow$	$controls_2$ at_1 $\neg confined_1$	9

Example State Transition Rules for the Aladdin Domain.

- Rules take the form E ⇒ S → F where:
  - E is a set of properties that enable the transition
  - S are the properties given up by the transition
  - F are properties acquired by the transition.
- The negative of a property x, used for analysing negative preconditions, is shown as ¬x.

Anton Process

- Antonym selection for the action marry
- The weights for each antonym are listed
- They are summed and the highest ranked is selected.
- In this case divorce is selected

Linguistic Resources	Antonyms for "marry"		
	dissociate	**divorce**	separate
Merriam-Webster	1	**2**	1
BigHuge Thesaurus	0	**0**	0
Power Thesaurus	1	**3**	1
Totals	2	**5**	2

① Action from Missing Transitions

(:action **marry**
:parameters (?m-male ?p-princess)
:precondition (and
(single ?m) (single ?p)
(loves ?m ?p) (loves ?p ?m) ...)
:effect (and
A (married ?m ?p) (married ?p ?m)
B (not (single ?m)) (not (single ?p))))

(:action divorce
:parameters (?m-male ?p-princess)
:precondition (and
(married ?m ?p) (married ?p ?m) A
(motivated-to-divorce ?m ?p) ...)
:effect (and
(single ?m) (single ?p) B
(not (married ?m ?p)) ...) A

② Action from Missing Properties

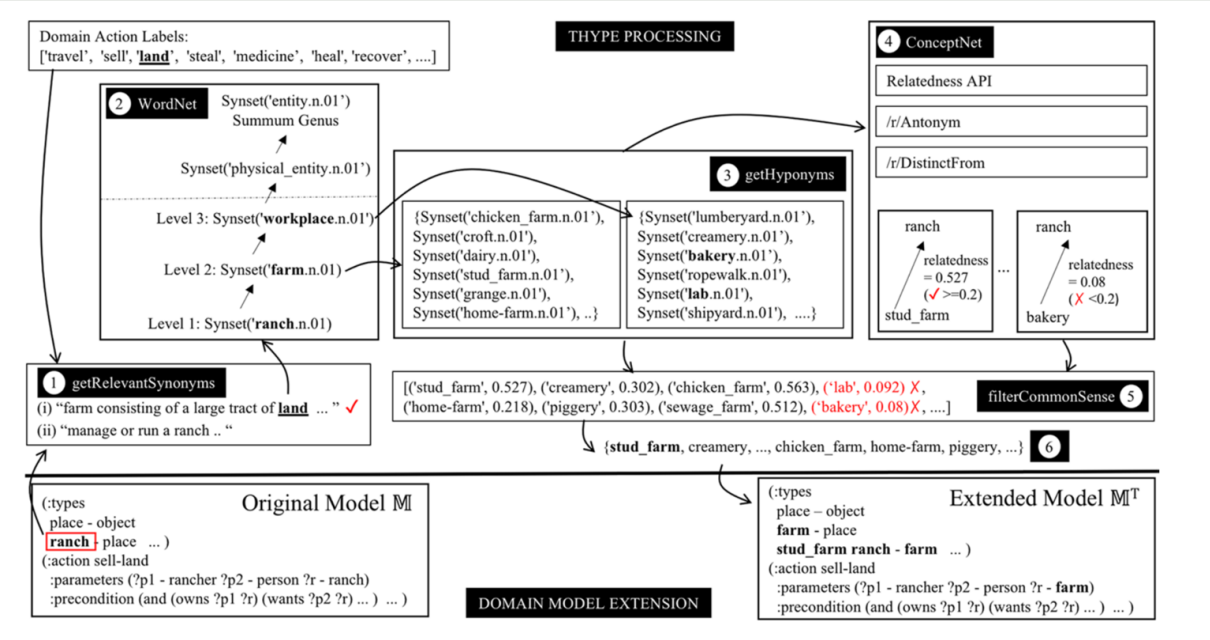
:action **become-beautiful**
:parameters (?p - princess)
:precondition (and (not (beautiful ?p)))
:effect (and (beautiful ?p)))

(:action **become-ugly**
:parameters (?p - princess)
:precondition (and (beautiful ?p))
:effect (and (not (beautiful ?p))))

③ New Predicates: e.g. replace (not (beautiful ?p) with (ugly ?p)

(:action **become-beautiful**
:parameters (?p - princess)
:precondition (and **(ugly ?p))**
:effect (and (beautiful ?p) (not **(ugly ?p)**))))

- New actions are generated using the original action(marry) as a template for the newly generated action (divorce)
- The pre-conditions are those predicates achieved by the action A (green circled) plus an enabling condition; the positive effects are the pre-conditions of the initial action B (green circled)



- takes an existing type from the original story domain (like ranch) and finds related types (like stud_farm, creamery, chicken_farm) to increase narrative diversity.

Original Model $\mathbb{M}$	Extended Model $\mathbb{M}^T$
(:types person - object king - person ...)	(:types ① person - object *sovereign* - person king *emperor* - *sovereign* ...)
(:action order :parameters (?k - king ...)	(:action order ② :parameters (*?k* - *sovereign* ...)

THYPE IN ACTION

Why does this matter?

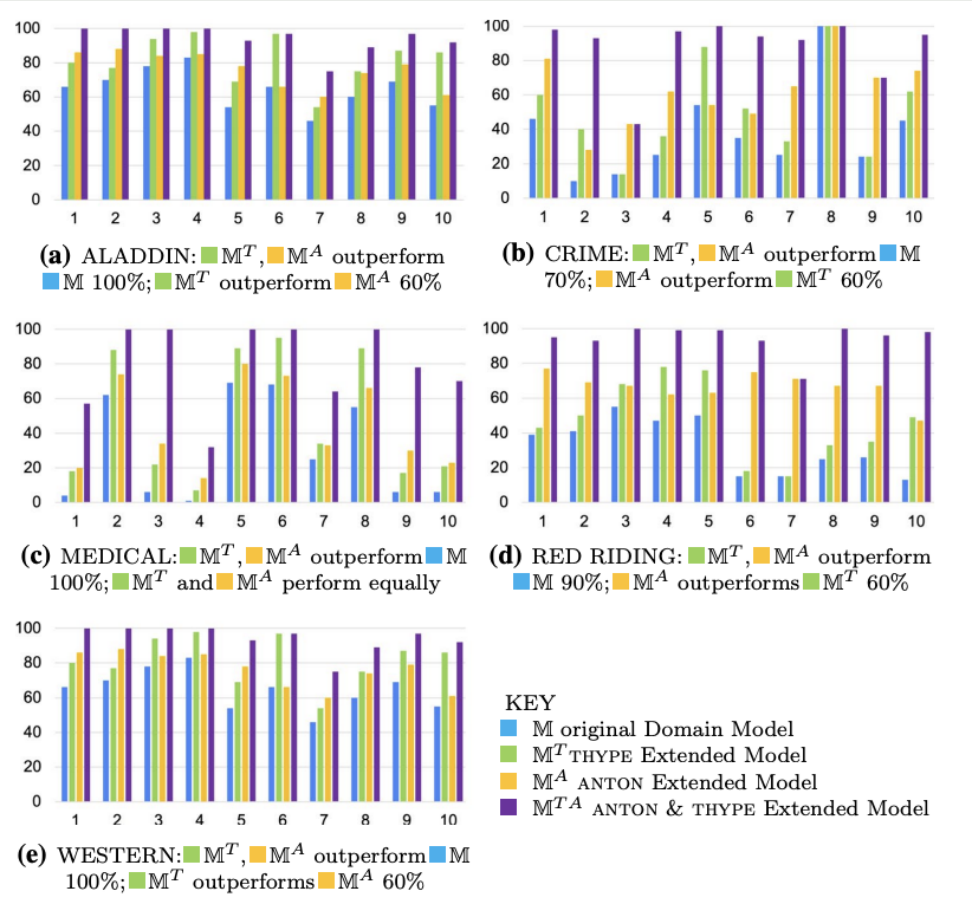
- Diversity: More possible storylines leads to better replayability
- Robustness: If a user breaks the story, planner can still reach an ending
- Human-readability: All actions/types are labeled using real language

Evaluation

- Tested on a range of narrative planning domains: Aladdin, Crime Drama, Medical Drama, Red Riding Hood, Western
- M, M^T, M^A, M^{TA}

Domain	Original Model $\mathbb{M}$ #A	#T	ANTON Model $\mathbb{M}^A$ #A	THYPE Model $\mathbb{M}^T$ #T
Aladdin	12	12	21	18
Crime	9	8	11	12
Medical	10	18	19	20
Red	5	10	9	14
Western	19	10	25	14

- We can observe that new actions and types were generated across all domains



- Robustness Results: comparing ability of (green square)T, (yellow square) A, (purple square) TA to continue to original goal in execution failure simulation (5 domains, 10 problems, 100 fail-restarts on each)
- M(skyblue square) is included for comparison.
- Performance of T and A are consistently good across all domains and outperform throughout.
- TA outperforms all other models and demonstrates the performance gains that can result from combining extensions

Original Content	THYPE content	Definition	Rank
Aladdin			
dragon	troll	a supernatural creature supposed to live in caves or mountains	G
dragon	werewolf	a monster able to change appearance from human to wolf and back	U
Crime Drama			
car	jeep		G
car	horseless-carriage		P

- Examples of thype generated content and user rankings (G(ood), P(oor) and U(nsure)).
- For the type “dragon” (Aladdin), the alternative type, “troll” was ranked G, whereas the alternative, “werewolf”, was ranked U
- for “car” (Crime Drama) the type “jeep” was ranked G, and “horseless-carriage” P.
- These rankings are genre dependent and can differ across domains e.g. a werewolf is plausible in Harry Potter but not Aladdin.

Strengths

- Automates a major bottleneck in narrative AI development
- Generalizable to many story genres
- Human-readable output makes adoption easier
- Demonstrates measurable gains in robustness & diversity

Weaknesses

- Relies heavily on external linguistic tools and dictionaries which could lead to error
- Some generated actions/types are not appropriate for narrative
  - Requires human filtering (not 100% automatic)

Conclusion

- This paper presents an automated way to enhance narrative AI planning models
- It significantly improves diversity and robustness

Thank You!