

Augmented Intelligence in Test Development: Using Constraint Programming for Item Generation

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Overview

- Automatic Item Generation has the ability to produce large numbers of items
 - Programming or instruction-based approach
 - Logical constraint and cognitive model-based approach (us!)
 - Ontology-based or knowledge-base dependent approach (e.g., Mitkov, Gutl or Soltadova)
- Our approach relies on combinatorial process with constraints to assemble a set of allowable combination of elements to represent a test item
- Focus of this talk is limited to how items are generated, for calibrated item generation...



Agenda

1. Where we come from

- Item Generation 101

2. How we do it

- Evolution of how we generate
- Augmented Intelligence: Constraint Programming

3. Applications

- English language items

4. Where we are going





Item Generation 101 – Rationale

The need for items

- Computer-based testing, adaptive testing
- Continuous administration
- Formative assessment
- Competency-based assessment

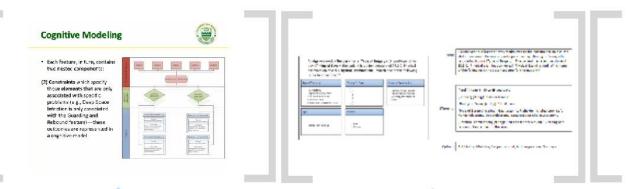
Scalability problem

- SMEs producing items one at a time
- Iterative review process (time consuming)
- Quality and guidelines eliminate items before reaching field test
- Need for items are increasing drastically while production is scaled by the number of SMEs





Our Approach





Step 1 Cognitive Modeling Step 2 Item Modeling Step 3 Item Generation



An example in surgery

A 54-year-old woman has a laparoscopic cholecystectomy. On post-operative day 3 she has a temperature of 38.5c. Physical examination reveal a red and tender wound and calf tenderness. Which one of the following is the best next step?

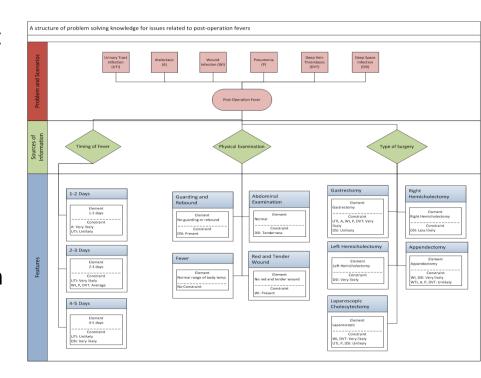
- a. Mobilize
- b. Antibiotics
- c. Anti coagulation
- d. Reopen the wound



Cognitive Modeling

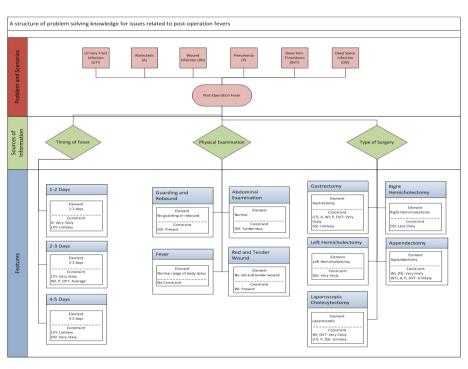
The model includes three key outcomes:

- Identify **THE PROBLEM** (i.e., Post-Operative Fever);
- 2. Specify **SOURCES OF INFORMATION** required to diagnose the problem (i.e., Type of Surgery); and
- 3. Describe **KEY FEATURES** within each information source (e.g., Fever) needed to create different instances of the problem





Item Modeling



A 54-year-old woman has a lapartscopic shall be still a postative day 3 she has a temperature of 38.5c. Physical examination reveal a perfect temperature of 38.5c. Physical examination reveal examination reveal

<**TYPE OF SURGERY>** Gastrectomy; Right Hemicolectomy; Left Hemicolectomy; Appendectomy; Laparoscopic Cholecystectomy

<TIMING OF FEVER> 1 to 6 days

<PHYSICAL EXAMINATION > Red and Tender Wound; Guarding and Rebound; Abdominal Tenderness; Calf Tenderness



Item Generation

- After the item model is specified, we combine this information systematically to produce new items
- To accomplish this combinatoric task, we use a software for item generation called IGOR (Item GeneratOR)
- IGOR was programmed using Sun Microsystems JAVA



- 1. A 54-year-old woman has a gastrectomy. On post operative day 3 he has a temperature of 38.5 C. Which one of the following is the most likely diagnosis?
- a. Urinary tract infection
- b. Actelectasis
- c. Wound infection
 d. Pneumonia
- 2. A 54-year-old woman has a gastrectomy. On post operative day 3 he has a temperature of 38.5 C. Which one of the following is the most likely diagnosis?
- a. Urinary tract infection
- b. Actelectasis
- c. Wound infection
- d. Deep vein thrombosis
- 3. A 54-year-old woman has a gastrectomy. On post operative day 3 he has a temperature of 38.5 C. Which one of the following is the most likely diagnosis?
- Urinary tract infection
- b. Actelectasis
- c. Wound infection
- d. Deep space infection
- 4. A 54-year-old woman has a gastrectomy. On post operative day 3 he has a temperature of 38.5 C. Which one of the following is the most likely diagnosis?
 - a. Urinary tract infection
 - b. Actelectasis
 - c. Pneumonia
- d. Wound infection



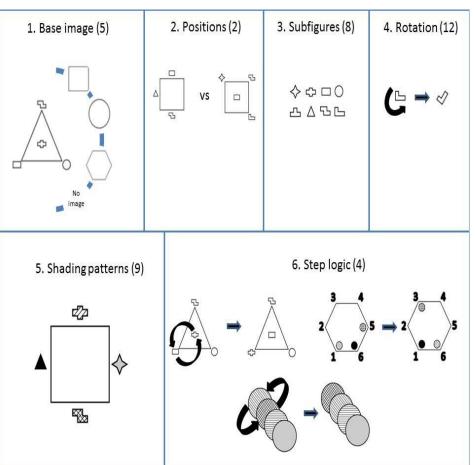
From that process

Mathematics (k-12)

Lai, H., Alves, C. & Gierl, M. (2009) Lai, H., Gierl, M., Hogan, J. (2015) Gierl, M. & Lai, H. (2012)

That's great, but what about...?

Topic Area	R
Medicine	Gierl, for mu Medic
Biology	Gierl, for Cc meeti
Non-verbal Reasoning	Gierl, Methc mode Intern Comn Spring
Dentistry	Lai, H (2016 Gene
Science	Gierl, press K-12: (Eds.) skill d



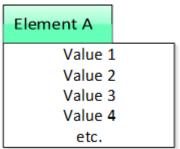
Evolution of how we generate

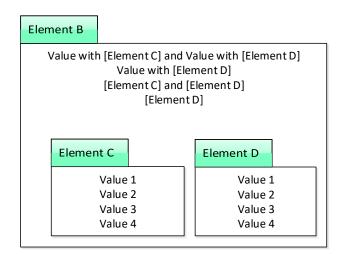
- Application of the three steps has allowed us to generate items
- As we moved into different domains, task of generating items became more complex
- We developed different tools to simplify the task of translating item models for generation



Tool 1. N-Layering

- N-layer permits the manipulations of a non-linear set of generative operations using a large number of elements at a multiple levels
- The concept of n-layer item generation follows development of syntactic structures for item creation (Bormuth 1969)
- Great for expressing item content in a different ways
- Think of this as increasing depth of a model







Tool 2. Linked elements

Element A.1

He She They

Element A.2

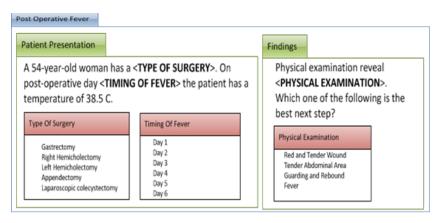
His Her Their

- Linking elements together allows manipulations of one set of elements to be dependent on another
- The concept of linking follows the idea of creating parallel content with a model
- Great for expressing consistent sets of content in items
- Think of this as increasing breadth of a model



Using these tools

By stacking various presentations in different layers and creating parallel elements in different languages...



	English	Chinese				
Sentence 1	Patient complaints of a mass	一名患者主訴				
	[[ENG.Pain]] in [[ENG.Location]]	[[CH.acuityofonsetCH]] 在				
	which has been a problem	[[CH.locationCH]] 出現的一個				
	since [[ENG.AcuityofOnset]].	[[CH.painCH]] 包塊。				
Sentence 2	A [[GEN.Gender]] was admitted	一名[[GEN.genderCH]]子因				
	with pain in [[ENG.Location]]	[[CH.acuityofonsetCH]][[CH.loc				
	from [[ENG.AcuityofOnset]].	ationCH]] 出現疼痛而入院。				
Sentence 3	On examination, the mass is	經檢查後,那包塊是				
	[[ENG.PhysicalFindings]] and	[[CH.physicalfindingCH]] 化 驗結				
	lab work came back with	果顯示[[CH.WBCch]]。				
	[[ENG.WBC]].					



Generating items in different languages

21. A 25-year-old woman presented with a mass in the left groin. It occurred a few months ago. On examination, the mass is protruding but with no pain and lab work came back with normal vitals. What is the best next step?

- 1. ice applied to mass
- 2. exploratory surgery
- 3. reduction of mass
- 4. hernia repair

47. 一名25歲的女患者在左側腹股溝出現一個包塊。徵狀己持續了幾個月。經檢查後, 那包塊是突出而不疼痛, 化驗 果顯示生命體徵正常。下一部最佳 處治是那一個?

- 1. 在包塊上冷敷
- 2. 腹腔探查術
- 3. 手法回復包塊
- 4. 疝氣修補手術

111. Patient presents with a mass and mild pain in an area near a recent surgery from a few days ago after moving a piano. The patient is a 25-year-old woman. Upon further examination, the patient had normal vitals and the mass is tender and reducible. What is the best next step?

- 1. ice applied to mass
- 2. exploratory surgery
- 3. reduction of mass
- 4. hernia repair

212. 一名患者的手術的切口附近從數天前, 自搬動鋼琴後出現一個有輕微痛感的包塊。患者牲別女, 25 歲。經身體檢查後, 患者生命體徵正常, 而那包塊是軟而可回復的。下一部最佳處治是那一個?

- 1. 在包塊上冷敷
- 2. 腹腔探查術
- 3. 手法回復包塊
- 4. 疝氣修補手術



Tool 3. Expression of Item Constraints

One of our biggest challenge in building item models was to program the constraints such that only plausible items will be generated

As models become more complex, expression of constraints turn into their own scripts

This moves our models back into programming approach for generation, and we need tools to move back into logical constraints



Old Constraints

Previously, constraints were expressed Boolean logic statements

A1	Apple	B1	tree
A2	Grape	B2	vine

```
[[A.value]] = = [[B.value]]
```

This was very time consuming when models become larger and relationships become more complex

```
 \begin{aligned} &(([[S5]] == 1) \&\& ([[11]] > 1)) \parallel (([[S5]] == 2) \&\& ([[11]] == 1)) \\ &(([[S6]] == 1) \&\& ([[12]] > 1)) \parallel (([[S6]] == 2) \&\& ([[12]] == 1)) \\ &(([[S7]] == 1) \&\& ([[13]] > 1)) \parallel (([[S7]] == 2) \&\& ([[13]] == 1)) \\ &(([[S8]] == 1) \&\& ([[14]] > 1)) \parallel (([[S8]] == 2) \&\& ([[14]] == 1)) \\ &(([[S9]] == 1) \&\& ([[11]]^*[[13]] > 1)) \parallel (([[S9]] == 2) \&\& ([[11]]^*[[13]] == 1)) \\ &(([[S10]] == 1) \&\& ([[12]]^*[[13]] > 1)) \parallel (([[S10]] == 2) \&\& ([[12]]^*[[13]] == 1)) \\ &(([[S11]] == [[S22]] \\ &(([[S11]] == [[S22]]) \&\& ([[S41]] == 2) \&\& ([[S42]] == 1)) \parallel (([[S11]] != [[S21]) \&\& ([[S41]] == 2)) \\ &(([[S1]] == [[S2]]) \&\& ([[S3]] == 2)) \parallel (([[S1]] != [[S2]]) \&\& ([[S3]] == 1)) \\ &([[S32]] == 2) \&\& ([[12]] != 1) \end{aligned}
```



New Constraint Programming

We moved into a new form of constraint programming that allows for the expression of relationships without using Boolean logic

For every pair of variables, there is a finite number of relationships of whether a value of one variable can be presented with another

If these relationships are captured in binary, then Boolean logic will not be needed

Models are just a values by values matrix, where at generation, a vector is called to determine where an item should be generated



An example of the new constraints

	В1	В2	В3	B4	B5
A1	1	1	0	1	1
A2	1	1	1	0	0
А3	0	0	0	1	0
A4	1	0	0	1	1
A5	1	1	1	1	1

Variable Pair Constraint Matrix

		Α	Α	Α	Α	Α	В	В	В	В	В	С	С	С	С
Varia	ble	1	2	3	4	5	1	2	3	4	5	1	2	3	4
Α	1	1	1	1	1	1	1	1	0	1	1	1	1	1	1
Α	2	1	1	1	1	1	1	1	1	0	0	1	1	1	1
Α	3	1	1	1	1	1	0	0	0	1	0	1	1	1	1
Α	4	1	1	1	1	1	1	0	0	1	1	1	1	1	1
Α	5	1	1	1	1	1	1	1	1	1	1	1	1	1	1
В	1	1	1	0	1	1	1	1	1	1	1	1	1	1	1
В	2	1	1	0	0	1	1	1	1	1	1	1	1	1	1
В	3	0	1	0	0	1	1	1	1	1	1	1	1	1	1
В	4	1	0	1	1	1	1	1	1	1	1	1	1	1	1
В	5	1	0	0	1	1	1	1	1	1	1	1	1	1	1
С	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1
С	2	1	1	1	1	1	1	1	1	1	1	1	1	1	1
С	3	1	1	1	1	1	1	1	1	1	1	1	1	1	1
С	4	1	1	1	1	1	1	1	1	1	1	1	1	1	1

Model Constraint Matrix

Given an item that uses A3,B5 and C3

Pá	air	Value		
А3	B5	0		
А3	C3	1		
B5	C3	1		
Outo	come	0		



New constraint programming benefits

Checkbox interface

No programming required

Lowers complexity of modeling

Much quicker generation time





Augmented Intelligence Application

Artificial intelligence is a term for describing the theory and the application of computer systems that perform tasks and solve problems that normally require human intelligence (Russell & Norvig, 2010)

Human-in-the-loop hybrid-augmented intelligence (or augmented intelligence, for short) is an AI topic that deals with how computer systems can extend human cognitive abilities thereby improving human task performance (Zheng, Liu, Ren, Ma, Chen, Yu, Xue, Chen, & Wang, 2017)



Augmented Intelligence Application

In the past, item modeling was created by the SME while the generation of items was done by programmers (workflow)

With the use of the new constraint programming approach, SME have the potential to naturally express constraints with the model and enable item generation without programming (direct manipulation)

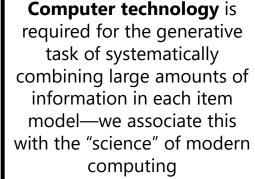
This is a crucial step towards enabling SME with augmented intelligence such that SMEs can improve their ability to generate items



Automatic item generation is becoming a symbiotic hybrid process where content specialists are being enhanced by technology to meet the demands of the producing test new test items

The **content specialist**identifies the knowledge
and skills required to solve
problems and the **test translator** expresses this
information across
language groups—we
associate these activities
with the "art" of test
development because it
requires judgement,
expertise, and experience

ART



SCIENCE

By merging the outcomes from the content-based creative task with the technology-based generative task, automated processes can be used to promote a new approach to item development

AIG





Generating English Language Arts Items

Reading Comprehension Item

At Home in a Language

When I was a child growing up in Delhi, India. My parents and I will have spent our summers in Calcutta, India, visiting my grandparents, aunts, uncles, and cousins. We took the train over eight hundred miles from Delhi to Calcutta, which I considered a treat as itself. I loved the dining car, the cozy sleeping berth in our cabin, and the gentle rocking motion of the train that would lull me to sleep at night. As an adult, I prefer to travel by car. When we arrived at the Calcutta station the next morning, we were welcomed announcing train arrivals and departures over the intercom by the sound of the Bengali language.

Back in Delhi, the language most people commonly spoke was Hindi. Though I spoke Hindi fluently, it wasn't my first language. My parents were born in Calcutta, when most people spoke Bengali. They had lived there for many years before they go married and moved to Delhi, where Hindi was widely spoken. Because my parents had grown up speaking Bengali, we spoke Bengali, not Hindi, in our house. It was not surprising, then, that hearing Bengali on the streets of Calcutta made me feel right at home.

Stem:

When I was a child growing up in Delhi, India. My parents and <u>I will have spent</u> our summers in Calcutta, India, visiting my grandparents, aunts, uncles, and cousins.



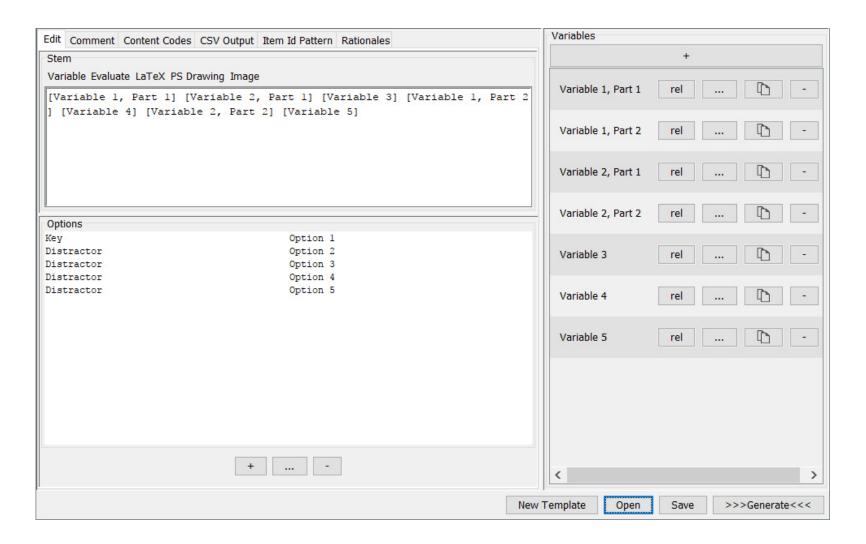
Item Model Stem:

[Variable 1, Part 1] [Variable 2, Part 1] [Variable 3] [Variable 1, Part 2] [Variable 4] [Variable 2, Part 2] [Variable 5]

Variable 1, Part 1	Variable 2, Part 1	Variable 3	Variable 1, Part 2	Variable 4	Variable 2, Part 2		Variable 5
When I was a child growing up in	Delhi, India,	my parents and	<u>I would spend</u>	our summers in	Calcutta, India,	visiting my	grandparents, aunts, uncles, and cousins.
During the year we lived in By the time I leave	Buffalo, United States, Cologne, Germany,	my family my siblings	<u>I spent</u> I will have spent	weekends holidays	Boston, United States, Berlin, Germany,		grandparents aunts and uncles
Were I to have stayed in	Tel Aviv, Israel,	my cousins	I would have spent	vacations	Jerusalem, Israel,		cousins
			<u>I spending</u>				
			<u>I will spend</u>				
			I will be spending				
			I have spent				
			NO CHANGE				







From 5832 possible combinations, 83 items generated



Generating English Language Arts Items

The same three step of AIG can be applied to generate items to meet the demands for ELA

New constraint programming enables generation of items in this area

While generation output is smaller in this area, there are lots of opportunities to modify and adapt the modeling approaches to fit the required assessment tasks

Future Directions

- Expansion the generative capacity in ELA
- Enhancing the quality control process of AIG
 - Developing guidelines that suit SMEs for AIG
 - Moving QC to the model level
- Further integrating SME within the process of AIG
 - Becoming more involved in the generation process
 - Integrating AIG into the item development workflow
 - Minimizing the outcomes from scalability



Summary

- Automatic Item Generation has the ability to produce large numbers of items
- Our approach relies on combinatorial process with constraints to assemble a set of allowable combination of elements to represent a test item
- With emergence of tools and aids that help SME, AIG is becoming an application of augmented intelligence in test development
- Methods for how items will generate items will continue to evolve, AI will continue to close the link between knowledge expression from SME and knowledge prompts for learners.



Thank you!

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