



Program Synthesis



MPhil ACS module P342 - Alan Blackwell



You do the rest!

Principles of program synthesis, from HCI perspective

- ▶ **The user experience of ML-based synthesis:**
 - ▶ The user says: “Here is an example of what I want to do”
 - ▶ Followed by: “You do the rest”
- ▶ **System response: “OK, I’ll do others the same way”**
 - ▶ How does it know what “others” are?
 - ▶ How does it know what “the same way” is?
- ▶ **Usability issues:**
 - ▶ How to specify applicability?
 - ▶ How to control generalisation?
 - ▶ How to understand what was inferred?
 - ▶ How to modify the synthesised program?

Classic programming by example

- ▶ Keyboard macros – demo in Emacs
- ▶ Get a plain text file containing semi-structured text
- ▶ **<Ctrl+x> (** starts macro recording
- ▶ Perhaps search for context, cut and paste, add text ...
- ▶ Remember to go to known location (e.g. start of next line)
- ▶ **<Ctrl+x>)** ends recording
- ▶ **<Ctrl+x> e** plays back once
- ▶ **<ESC> 1 0 0 <Ctrl+x> e** repeats 100 time

Value proposition

- ▶ The next generation of AI: “Intelligent tools”
- ▶ If a user knows how to perform a task on a computer, that should be sufficient to create a program to perform the task.
 - ▶ Early research aimed to achieve “programming in the user interface”
- ▶ Macro recorders are one model, but they are “too literal”
 - ▶ Do only what they are shown (no generalisation)
 - ▶ Unable to adjust for different cases (no inference)
- ▶ Other models:
 - ▶ Automation of repetitive activities
 - ▶ Creation of custom applications
- ▶ Machine learning problem is to create a model of user *intent*
 - ▶ Ideally informed by prior likelihood – from this user, and other users

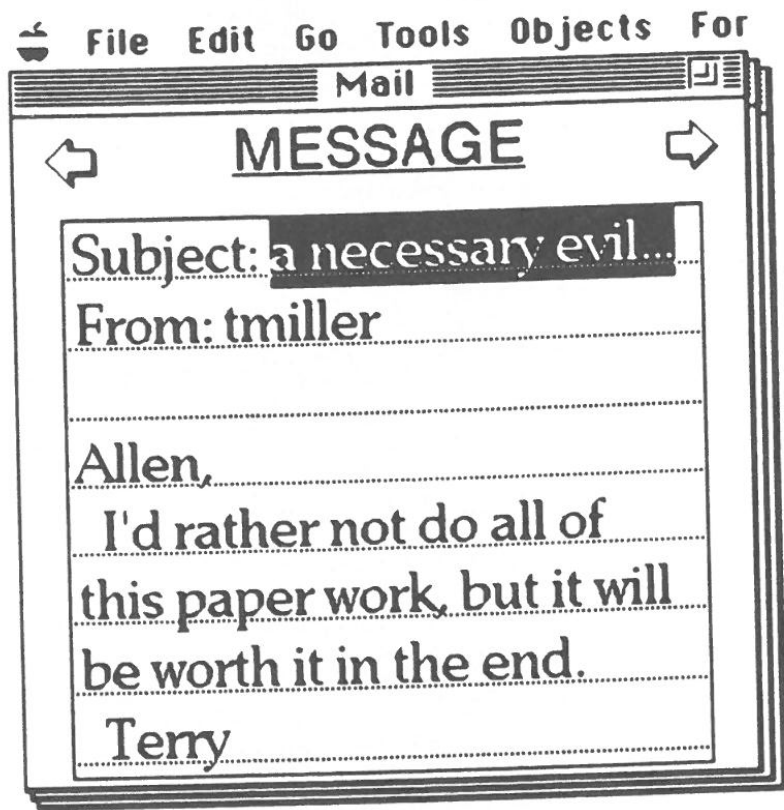


Eager

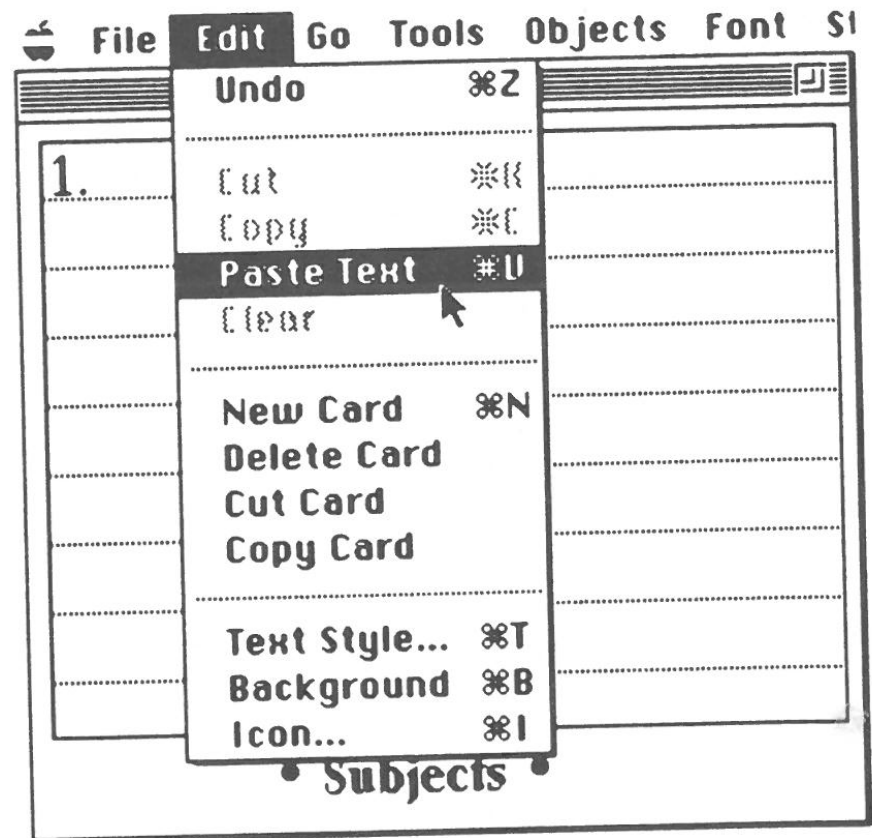
Classic mixed-initiative programming by example

- ▶ Allen Cypher's "Eager" created at Apple research in 1990
 - ▶ Implemented as extension to Hypercard (event capture + injection)
 - ▶ Machine learning implemented in LISP
- ▶ Scenario – create a script to produce a list of subject lines from messages

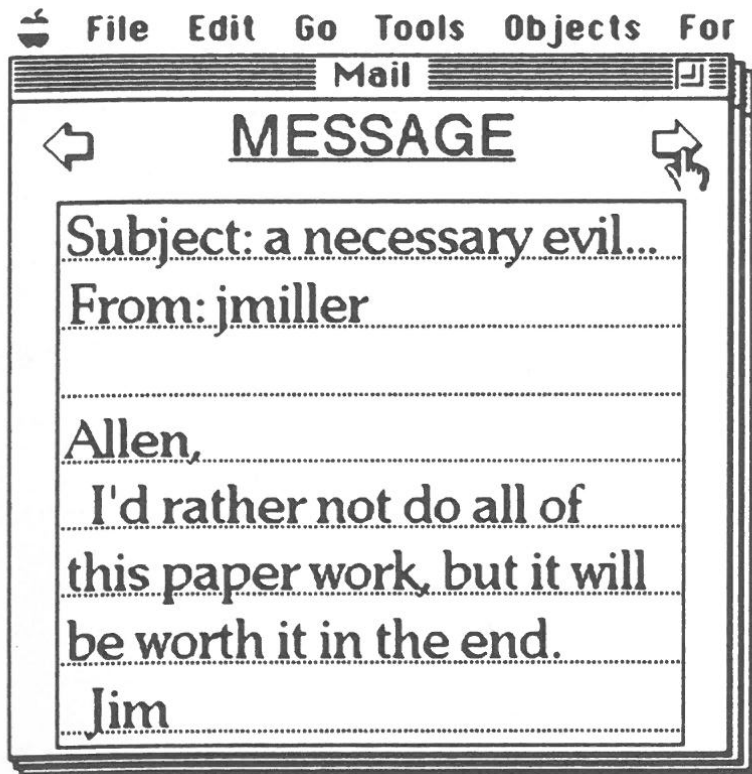




a) copy first subject



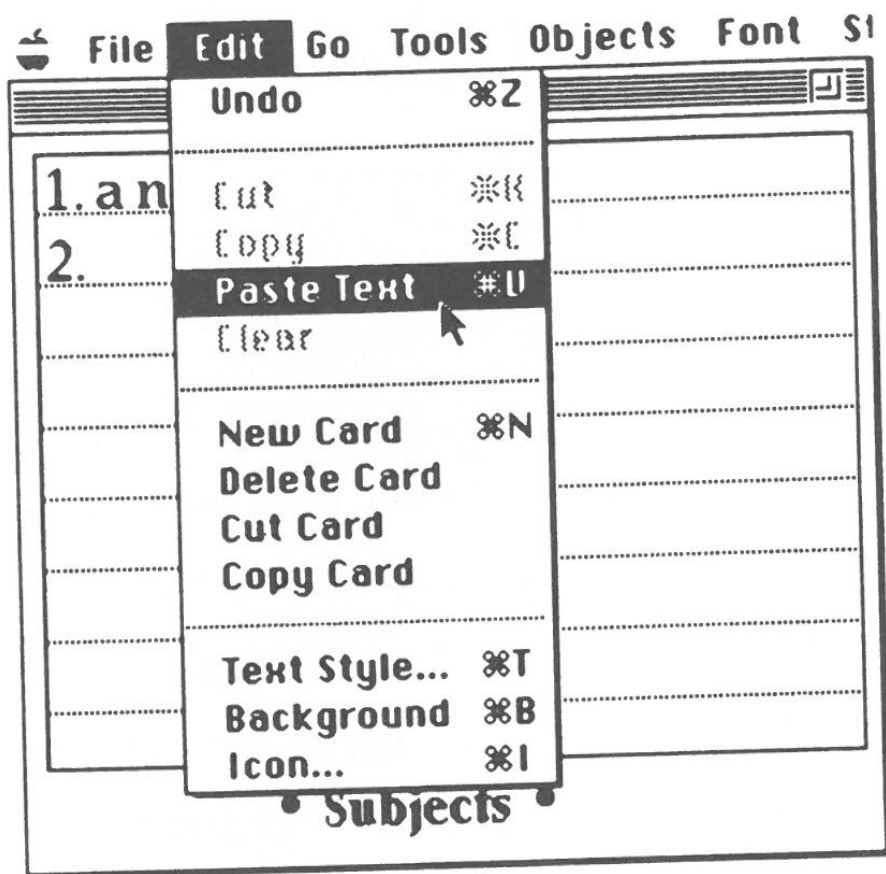
b) type "1. " and paste subject



c) go to next message



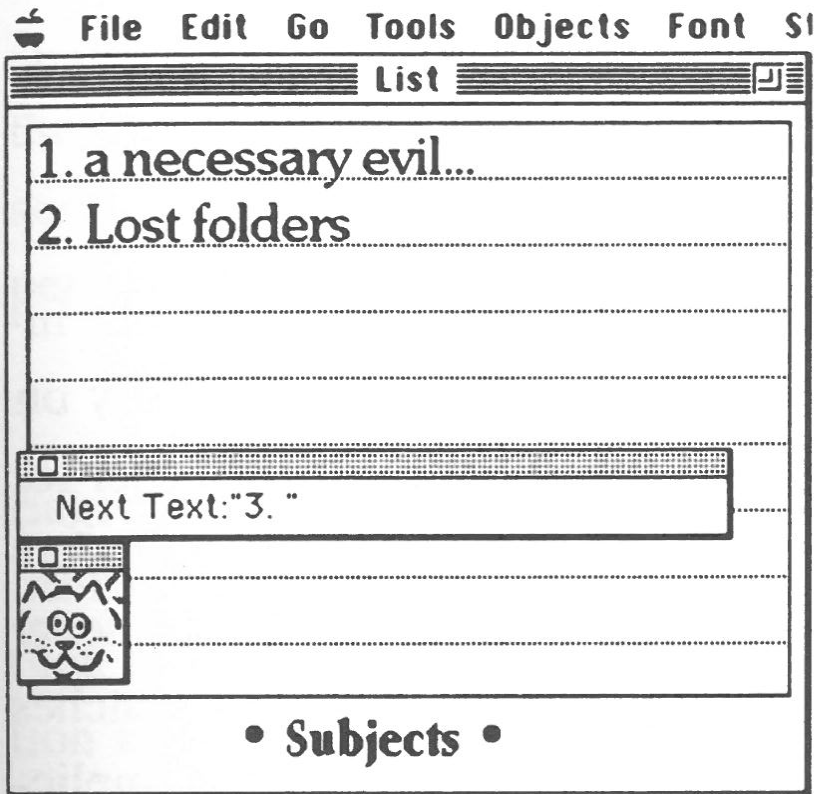
d) copy second subject



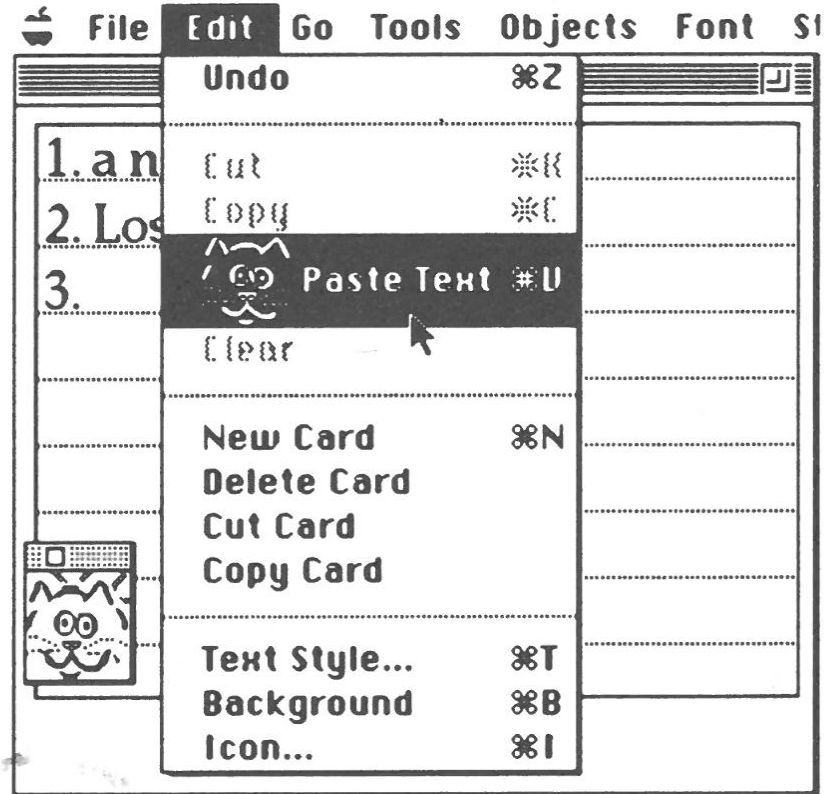
e) type "2. " and paste subject



f) Eager appears



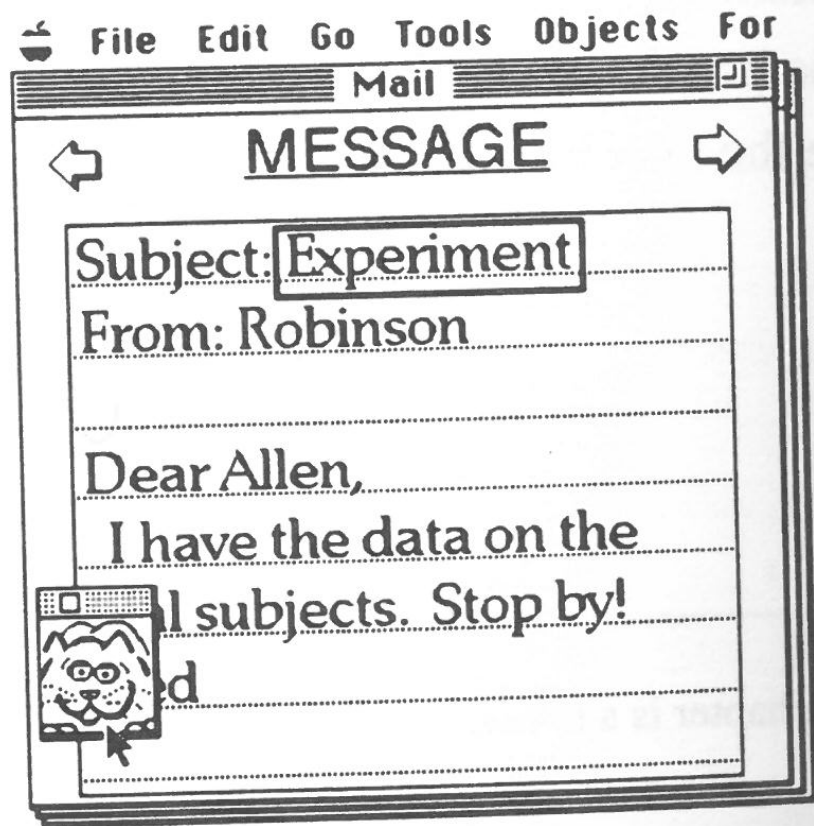
g) anticipate typing "3. "



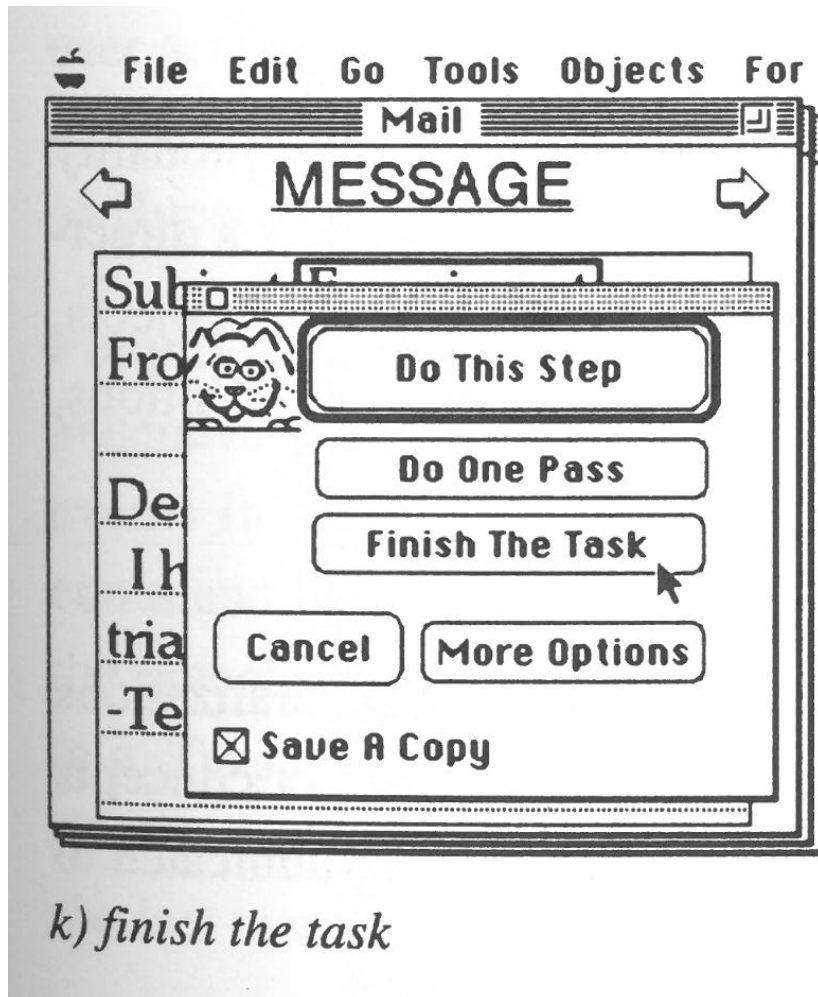
h) anticipate paste



i) anticipate going to next message



j) user clicks on Eager



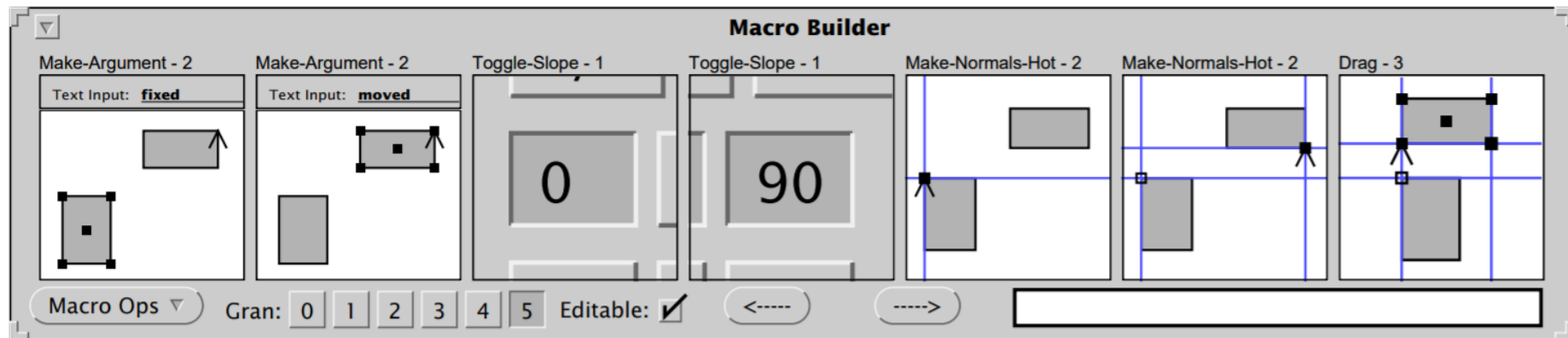


Chimera

Programming by demonstration in the graphics domain

- ▶ **Classic example: David Kurlander's Chimera**

- ▶ Infers constraints via heuristics, from snapshots of drawing editor state
- ▶ Users can generalise a “graphical macro” in editable history of operations
- ▶ <https://youtu.be/JbrjQW25ekl?t=7m7s>



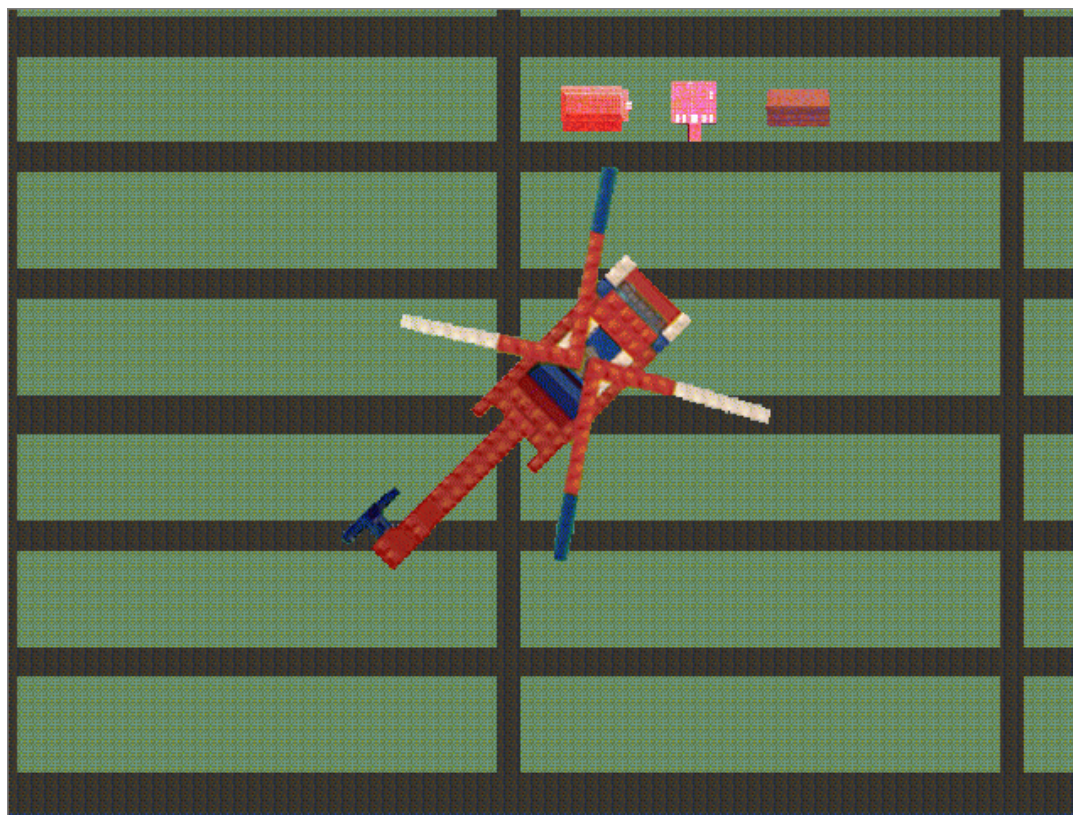
- ▶ **D. Kurlander *Graphical Editing by Example* (1993)**

- ▶ PhD thesis, Columbia University. CS Tech/ Report CUCS-023-93

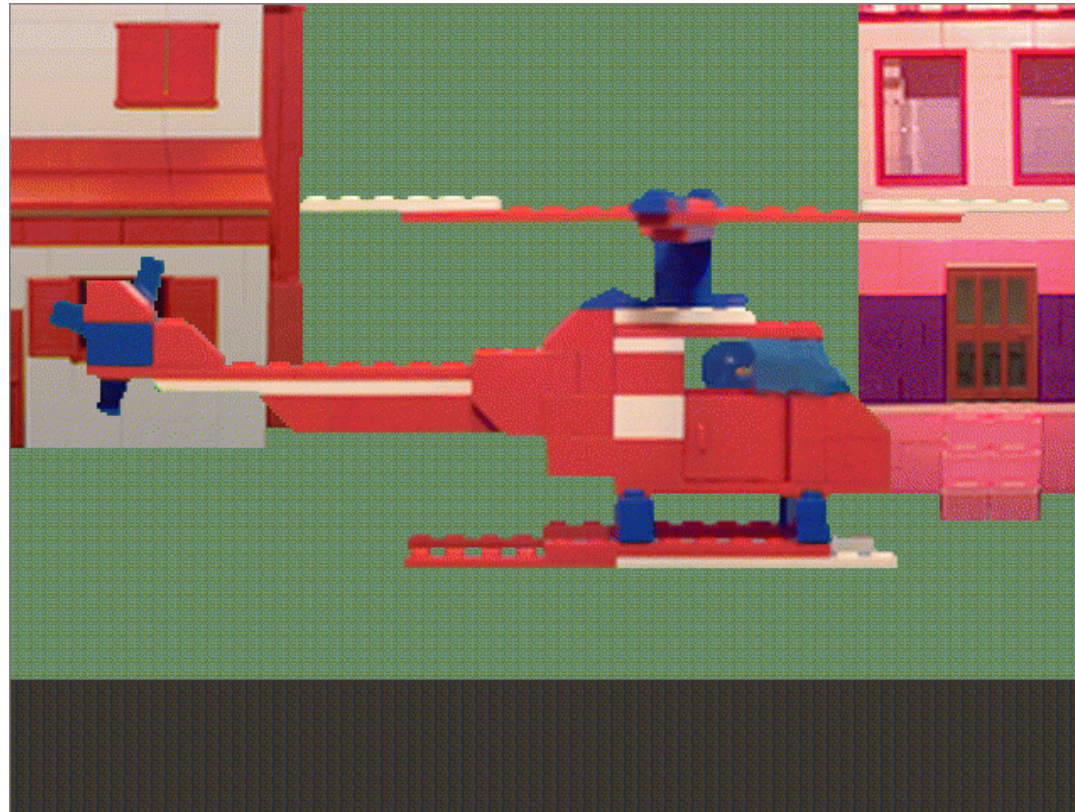


ToonTalk

Ken Kahn's ToonTalk – user control of generalisation



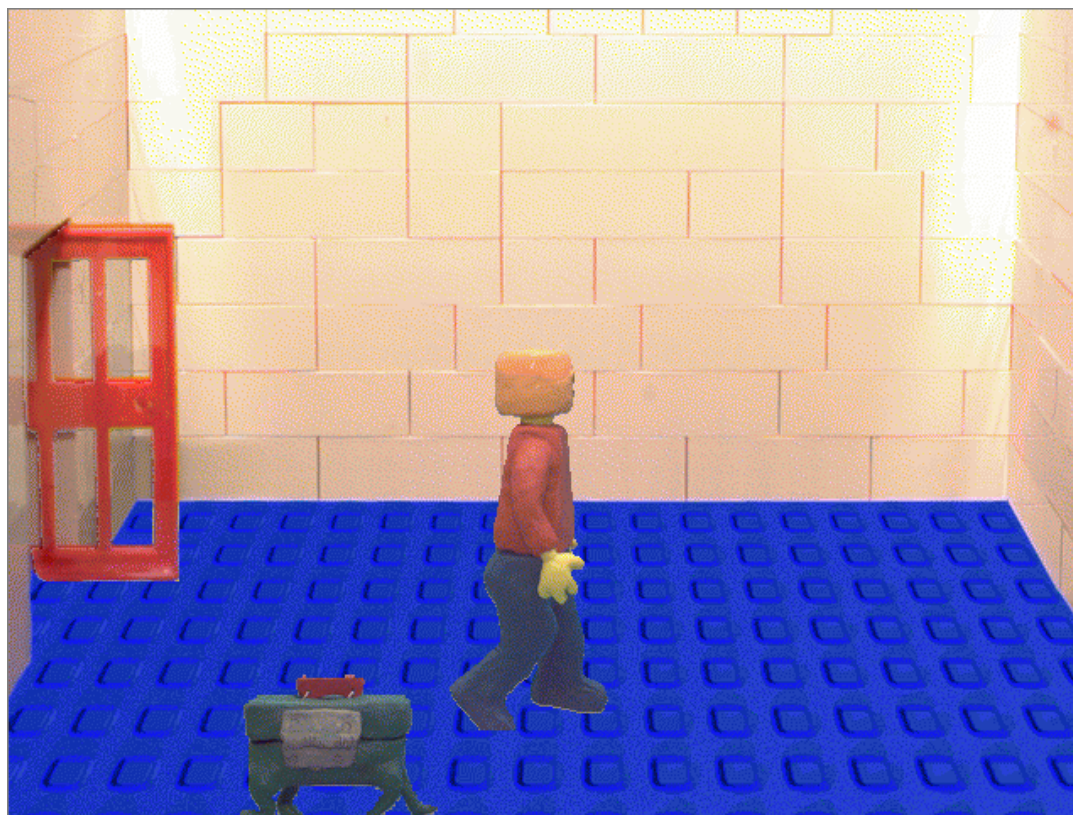
Ken Kahn's ToonTalk – user control of generalisation



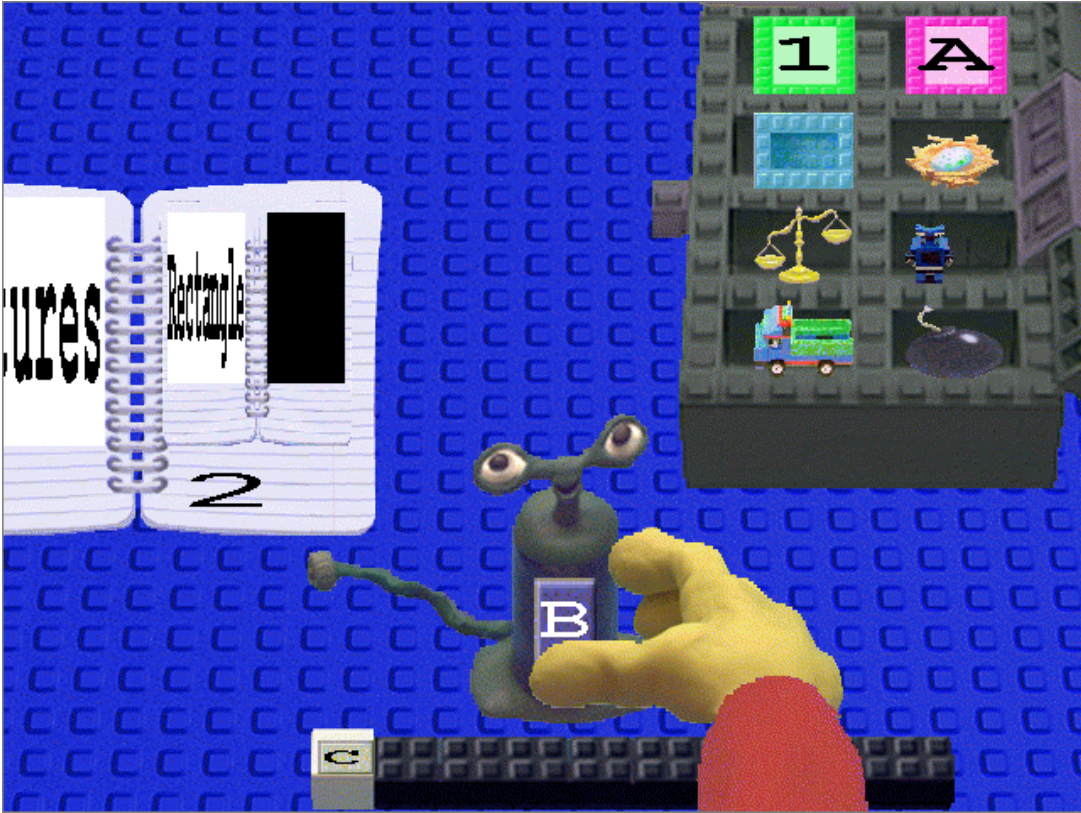
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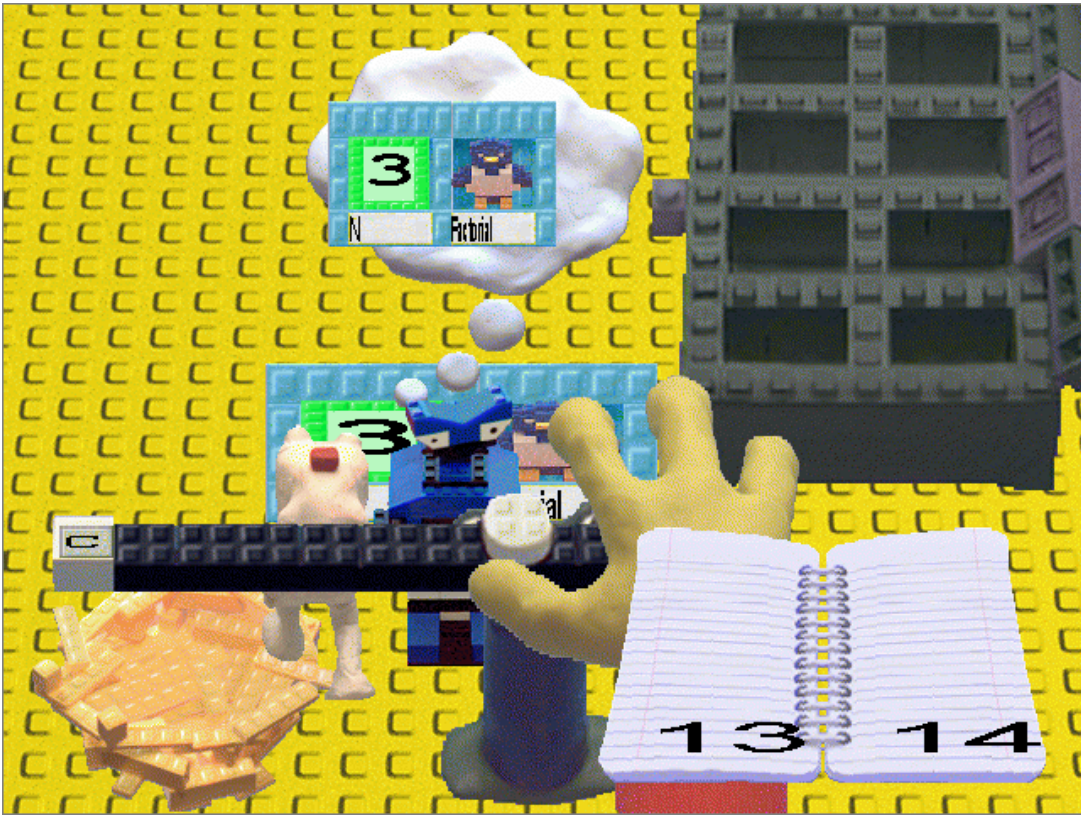
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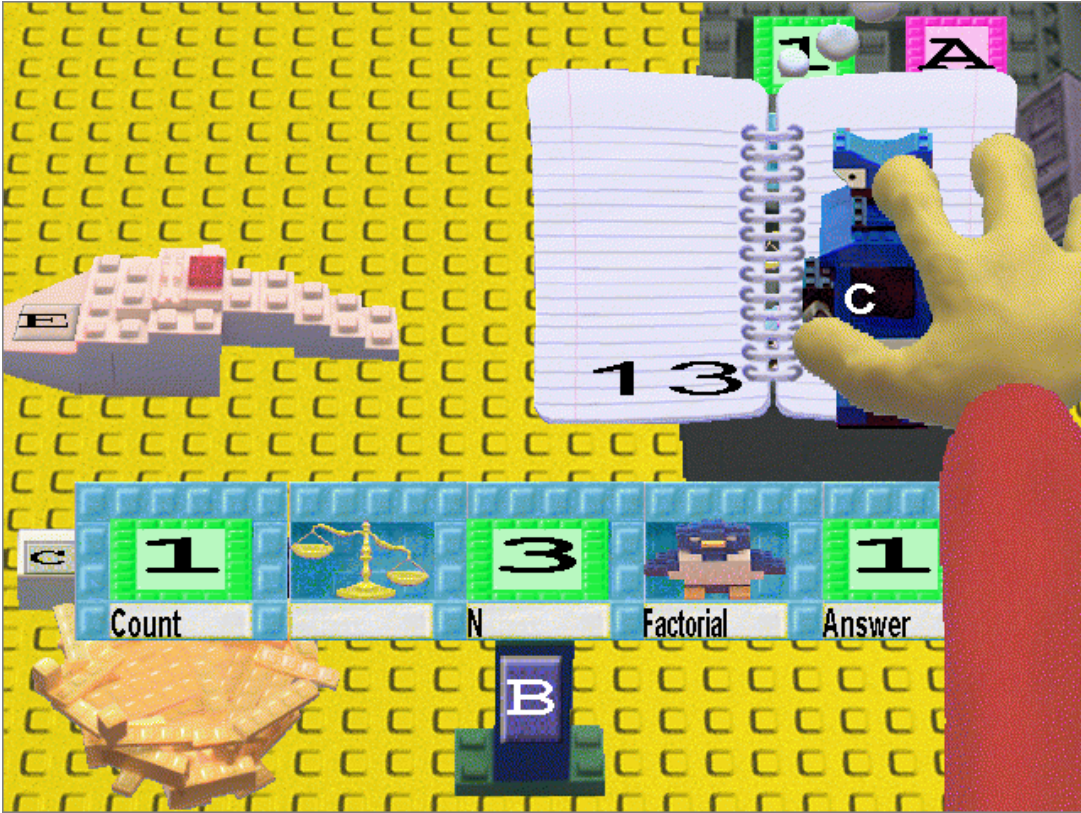
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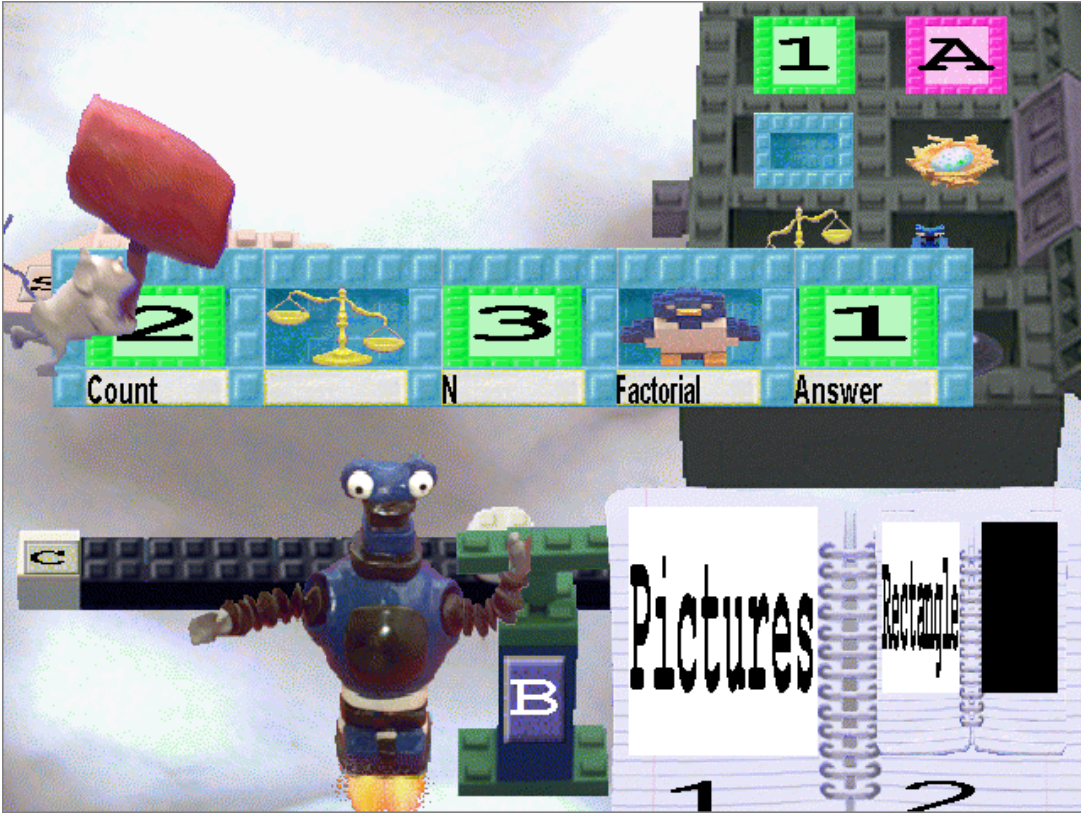
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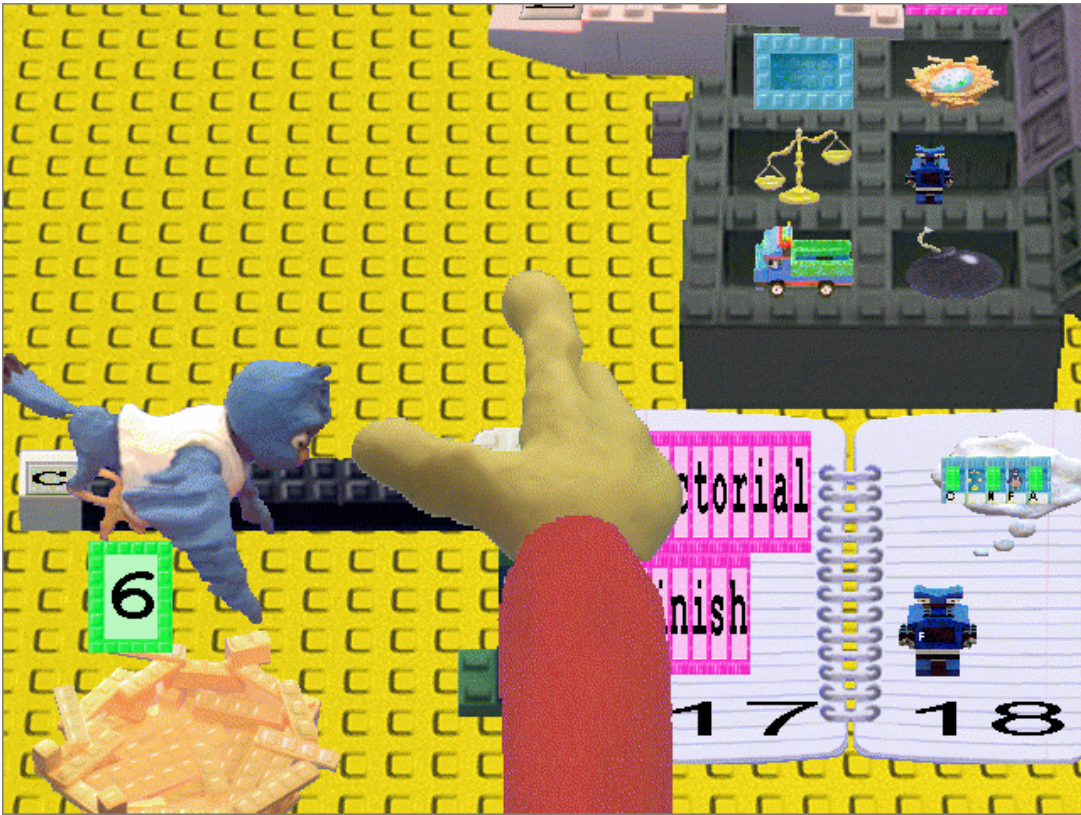
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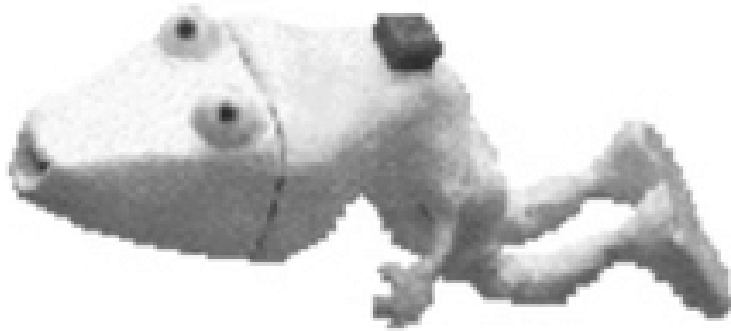
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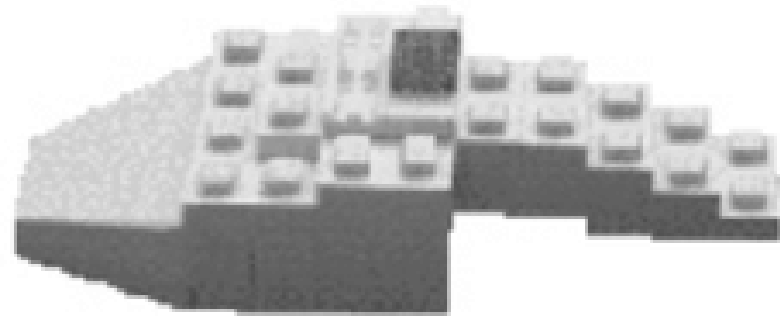
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Generalising a constraint with Dusty



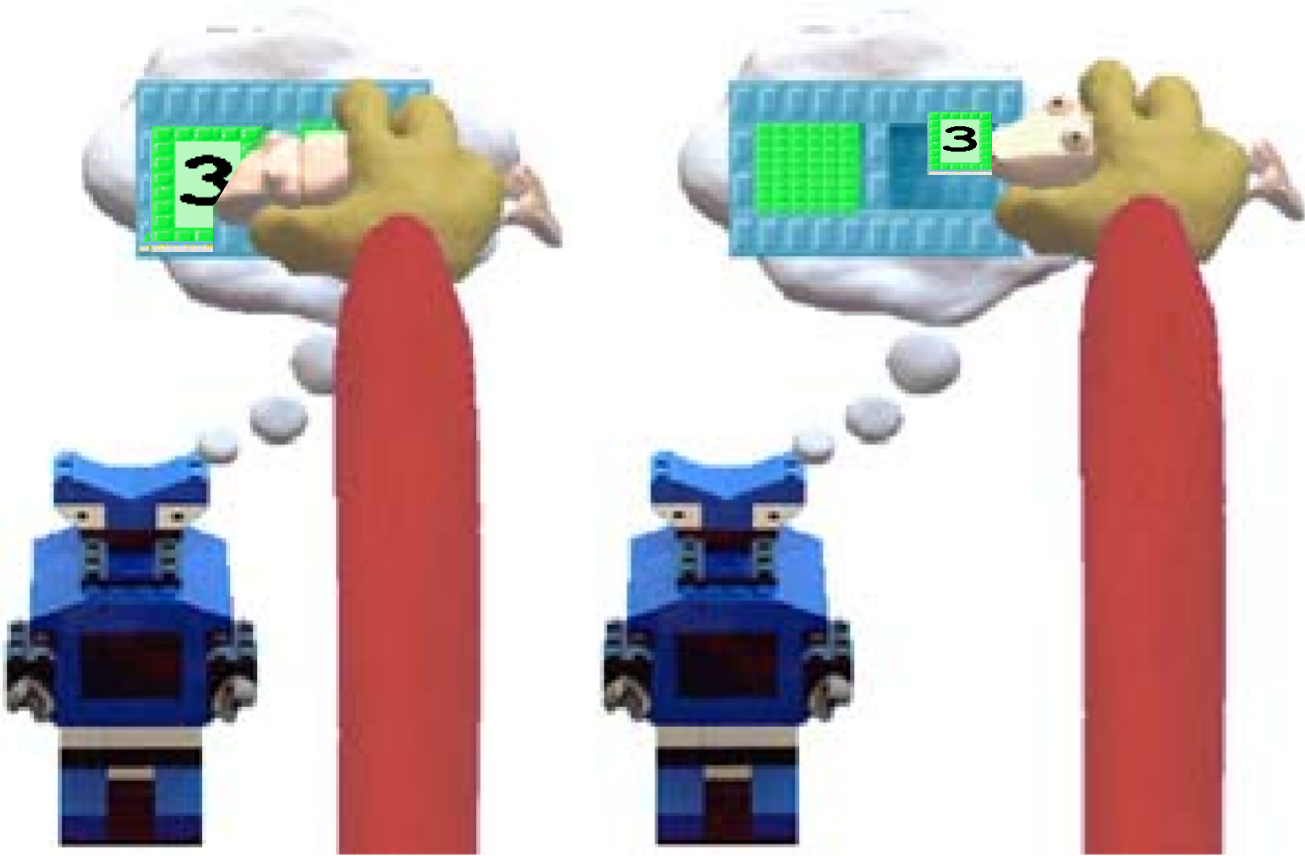
Dusty (a)



Dusty (b)



Generalising a constraint with Dusty





Generalisation

Why is the generalisation step so significant?

- ▶ **Generalisation from examples is fundamental to mental abstraction**
 - ▶ Repetition of concrete instances (i.e. direct manipulation) does not require abstraction
 - ▶ Any automated action (i.e. programming) does require abstraction
- ▶ **So program synthesis requires the user to conceptualise their problem in an abstract way**
 - ▶ Programming by example is a strategy for achieving this ...
 - ▶ ... the user can become comfortable with individual cases, while
 - ▶ ... the system formulates abstractions at the same time the user does.
- ▶ **Essential that user & system can “discuss” what they are concluding:**
 - ▶ So is this what you want me to do?
 - ▶ No, here is a case where you should do something else.
 - ▶ Oh, I see, so like this?

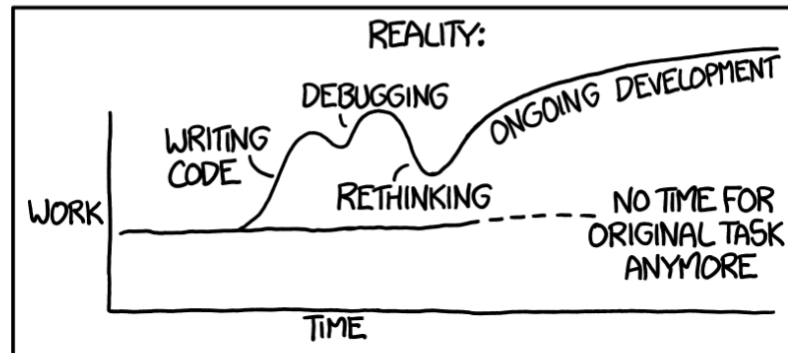
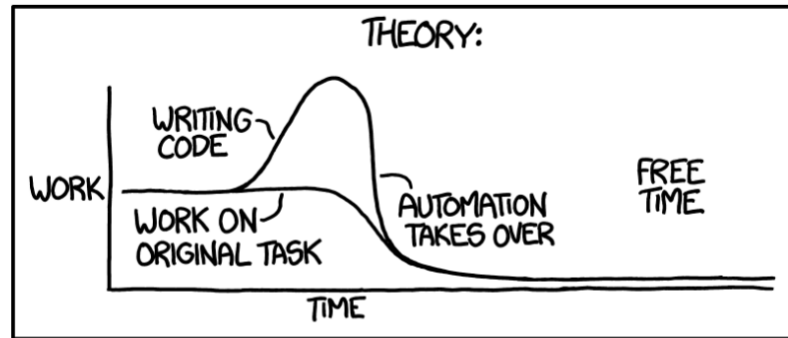
The Attention Investment model of abstraction use

- ▶ Programming is not like direct manipulation, so the standard rules of usability (Shneiderman's direct manipulation principles) do not apply:
 - ▶ Incremental action
 - ▶ Fully visible state
 - ▶ Immediate feedback
 - ▶ Easily reversible actions
- ▶ Making abstractions is cognitively hard, because *actions take place in the future*, and they apply to *multiple potential contexts*.
 - ▶ Automating repetitive actions does save time and (mental) effort
 - ▶ But formulating and refining abstractions costs time and mental effort!
 - ▶ What leads a user to approach their tasks in this way?
 - ▶ Richard Potter's "Just In Time Programming"
 - ▶ Rosson and Carroll's "Paradox of the Active User"
 - ▶ Bainbridge's "Ironies of Automation"
 - ▶ Burnett's "Surprise, Explain, Reward" (cf mixed-initiative design strategies, including Clippy)

AUTOMATION

|< < PREV RANDOM NEXT > >|

"I SPEND A LOT OF TIME ON THIS TASK.
I SHOULD WRITE A PROGRAM AUTOMATING IT!"



|< < PREV RANDOM NEXT > >|

PERMANENT LINK TO THIS COMIC: [HTTPS://XKCD.COM/1319/](https://xkcd.com/1319/)

IMAGE URL (FOR HOTLINKING/EMBEDDING): [HTTPS://IMGS.XKCD.COM/COMICS/AUTOMATION.PNG](https://imgs.xkcd.com/comics/automation.png)



SWYN: See What You Need

Swyn: inferring regexps to generalise text macros

wibble wobble tries to nobble
~~wibbre~~ wobble tries to nobble
wibble wubbse tries to nobble
wibble wobble tries to nobble
wibble wobble tries to nobble
wibble wobble tries to nobble
wibble wobble tries to nobble
wibble wobble tries to nobble
wibble wubble tries to nobble
wibbbbbbble tries to trouble
wibbne wobble tries to nobble

Swyn: inferring regexps to generalise text macros

wibble wobble tries to nobble
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wibble ~~wubbe~~ tries to nobble
wibble wobble tries to nobble
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wibble wobble tries to nobble
wibble wobble tries to nobble
wibble wubble tries to nobble
wibbbbbbble tries to trouble
~~wibbne~~ wobble tries to nobble

Swyn: inferring regexps to generalise text macros

wibble **wobble** tries to nobble
~~wibbre~~ **wobble** tries to nobble
wibble ~~wubbse~~ tries to nobble
wibble **wobble** tries to nobble
wibble **wobble** tries to nobble
wibble **wobble** tries to nobble
wibble **wobble** tries to nobble
wibble **wobble** tries to nobble
wibble ~~wubble~~ tries to nobble
wibbbbbbble tries to trouble
~~wibbne~~ **wobble** tries to nobble

Swyn: inferring regexps to generalise text macros

wibble wobble tries to nobble
~~wibbre~~ ~~wobble~~ tries to nobble
wibble ~~wubbse~~ tries to nobble
wibble wobble tries to nobble
wibble wobble tries to nobble
wibble wobble tries to nobble
wibble wobble tries to nobble
wibble ~~wubble~~ tries to nobble
wibbbbbbble tries to trouble
~~wibbne~~ wobble tries to nobble

Swyn: inferring regexps to generalise text macros

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wibble wobble tries to nobble
wibble wobble tries to nobble
wibble wobble tries to nobble
wibble wobble tries to nobble
wibble ~~wubble~~ tries to nobble
~~wibbbbbb~~ tries to ~~trouble~~
~~wibbne~~ wobble tries to nobble

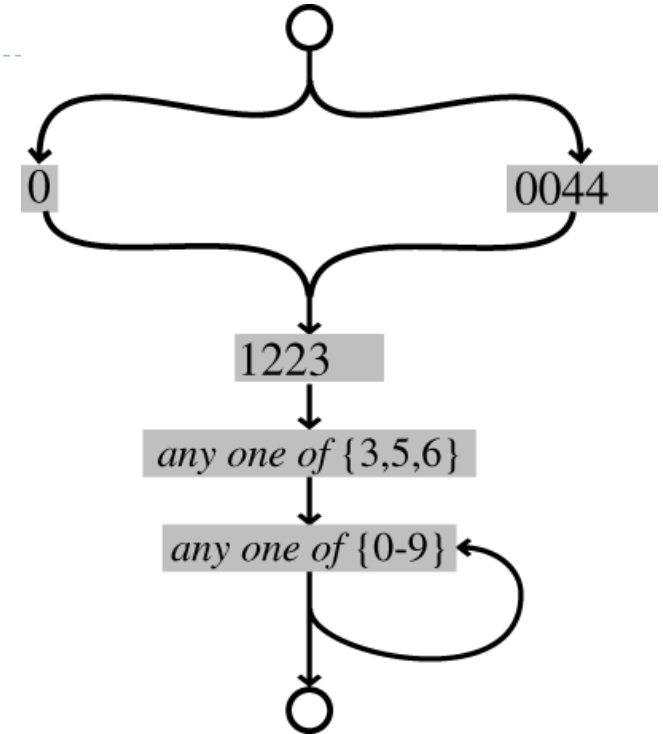
Swyn: inferring regexps to generalise text macros

wibble wobble tries to nobble
~~wibbre~~ ~~wobble~~ tries to nobble
wibble ~~wubbse~~ tries to nobble
wib ~~le~~ wobble tries to nobble
wib **trou** e tries to nobble
wib ~~le~~ wobble tries to nobble
wib **wob** **?** e nobble
wibble wobble tries to nobble
wibble ~~wubble~~ tries to nobble
~~wibbbbbb~~ tries to ~~trouble~~
~~wibbne~~ wobble tries to nobble

Explaining the inference to end-users

▶ (0|0044)|223[356][0-9]+

- ▶ Find one of the following:
 - ▶ a) either the sequence “0”
 - ▶ b) the sequence “0044”
- ▶ followed by the sequence “1223”
- ▶ followed by
 - ▶ any one of these characters: “3” or “5” or “6”
- ▶ followed by at least one, possibly more, of the following:
 - ▶ any one of these characters: any one from “0” to “9”



0
0044

1223

3
5
6

0-9

Structured text editing as an ML application

- ▶ Aimed at the kind of things people did with sed/awk/perl
 - ▶ Many automated text operations involved regexps
 - ▶ But users found these the hardest thing to understand ...
 - ▶ ... research agenda for machine learning: sed/awk/perl/swyn
- ▶ Similar goals to Witten and Mo's TELS (1989)
 - ▶ Learning Text Editing Tasks from Examples
 - ▶ See Cypher book chapter 8
- ▶ Luke Church demonstrated working solution (2007)
 - ▶ Recursive language model "Structured Prediction by Partial Match"
 - ▶ Prior expectation based on harvested corpus of regular expressions



Example applications

We've had intelligent code editors since 1978

- ▶ **The *Programmer's Assistant*: Knowledge-Based Emacs (KB-Emacs)**
 - ▶ PhD project of Charles Rich at MIT
 - ▶ Aimed to recognise cognitive plan elements within source code
- ▶ **However, most programmer-assist features in IDEs were implemented using context-aware heuristics rather than AI models:**
 - ▶ Syntax-directed editing
 - ▶ Auto-complete of standard constructs
 - ▶ Refactoring
 - ▶ Inference from identifier names (e.g. follow `x=x+1;` with `y=y+1;`)
 - ▶ Navigate-by-completion for library APIs
- ▶ **This is one area where LLMs (e.g. GitHub CoPilot) will have massive effects!**
 - ▶ Ask Advait. (Or Michael Lee?)

Working in a data-centric paradigm: FlashFill for Excel

- ▶ **Building on this paper by Sumit Gulwani (MSR Redmond)**
 - ▶ *Automating String Processing in Spreadsheets using Input-Output Examples*, Proc. POPL 2011
 - ▶ <https://www.microsoft.com/en-us/research/publication/automating-string-processing-spreadsheets-using-input-output-examples/>
- ▶ **Live Demo**
 - ▶ Paste a list of semi-structured text data into the left column
 - ▶ Type an example transform result in top cell to the right, then <Enter>
 - ▶ Press <Ctrl+E>
- ▶ **“Synthesises a program from input-output examples”**
 - ▶ How do you choose the examples?
 - ▶ How do you know what will happen?
 - ▶ Using this ‘program’ as a component of a larger system is still a research topic

Visualising abstract structure: Data Noodles

- ▶ <https://www.youtube.com/watch?v=hyCVBxfx7VE>
- ▶ Applies a transformation paradigm
 - ▶ Directed search for fold/unfold transforms that will achieve the demonstrated result
- ▶ Search procedure uses off-the-shelf program synthesis toolkit
 - ▶ PROSE SDK from Gulwani team at MSR Redmond
- ▶ Custom-built front-end
 - ▶ The “spreadsheet” is purely for familiarity of presentation
 - ▶ No actual spreadsheet calculation is performed
 - ▶ Drag-and-drop target previews allow user to anticipate inference
 - ▶ Noodles preserve and visualise the demonstrated actions
 - ▶ Allow reasoning about causality from example to synthesised program
 - ▶ Potentially support modification/correction of examples