# From State- to Delta-Based Bidirectional Model Transformations: The Symmetric case 10 years after 

Zinovy Diskin, Yingfei Xiong, Krzysztof Czarnecki, Hartmut Ehrig, Frank Hermann, Fernando Orejas

In memory of Hartmut Ehrig who passed away on March 17, 2016

Models'21 MIP Talk, Fukuoka, Oct 13, 2021

## How we got together

M. Antkiewicz \& K.Czarnecki, Design Space of Heterogeneous Synchronization @ GTTSE’07


Zinovy
Z.Diskin, Y.Xiong \& K.Czarnecki: From State- to Delta-Based Bidirectional Model Transformations: the Asymmetric Case @ ICMT'10


Krzysztof Waterloo, 2007-11 Dagstuhl, Dec 2010


Yingfei


Frank

Two papers at Models'11


Hartmut
F. Hermann, H. Ehrig, F. Orejas, U. Golas: Formal Analysis ...for Model Transformations Based on TGG @ ICGT'10

## Content

- Background
- Our contribution: Why deltas
- Sync in the large: Multi-ary delta lenses (mx) and lens composition
- Applications in DB, SE, PL
- Applications in ML and AI: looking forward


## Change propagation is everywhere

- Our world consists of interrelated objects
- Objects are subjected to changes that can violate consistency between objects
- Consistency is to be restored by change propagation
- Mathematical models of
 change propagations are often called lenses


## Typical Cases of Change Propagation



Database and its Views

GUI and its underline model



UML Diagrams and Code


Source and target programs

## Example


$: b P p g$


## State-based Bidirectional Transformation (Bx)

- Consistency Relation
- $R: A \times B$
- Forward Propgation
- fPpg : $A \times A \times B \rightarrow B$
- Backward Propagation
- bPpg : $A \times B \times B \rightarrow A$



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- Problems of state-based Bx
- Delta-based Bx and Laws
- Big Picture
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## Update Interpretation: Delta 1




Delta was called "vertical delta" in the paper.

## Update Interpretation: Delta 2




## Problem 1 of State-based BX

## State-based Change Propagation



- Semantics of DD is mixed into DP, complicating the latter


## Problem 2 of State-based Bx

State-based Change Propagation


- New deltas are discarded, causing composition problem


## Inconsistent Horizontal Composition



## Inconsistent Horizontal Composition



Efficiency: b2 has to compute the delta again Semantics: b2 may compute a different delta

## Correspondence Interpretation: Corr 1



Correspondence was called "horizontal delta" in the paper and was sometimes called "traceability" in other papers.

## Correspondence Interpretation: Corr 2



## Erroneous vertical composition



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## Delta-based Bx

## R: a set of deltas between two model spaces



## Benefits 1

- Semantics of delta discovery and propagation are separated - When state-based BX is needed, we can add a DD component.
- No composition problem

$$
\boldsymbol{A} \stackrel{r}{r} \boldsymbol{B}
$$



## Laws

- Laws define the behavior of $B x$

- Well-behaved Bx: satisfying the above laws


## Benefits 2

- Laws are needed to prevent unnecessary changes

- Delta-based Bx allows laws that are impossible for state-based Bx
- Weak undoability
- Weak invertibility


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## Asymmetric and Symmetric Bx



Symmetric Bx
Consistency is defined by a binary relation $R$


Asymmetric Bx
Consistency is defined by a function get

| Symmetric bx | Asymmetric bx |
| :--- | :--- |
| R | $(s, \operatorname{get}(s)) \in S \times V$ |
| fPpg | $\operatorname{get}: S \rightarrow V$ |
| bPpg | put: $S \times V \rightarrow S$ |
| Correctness | PutGet: $\operatorname{get}(\operatorname{put}(s, v))=v$ |
| Hippocraticness | $\operatorname{GetPut}: \operatorname{put}(s, \operatorname{get}(s))=s$ |

## History and Classification



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- Our contribution: Why deltas
- Sync in the large: Multi-ary delta lenses (mx) and lens composition
- Why Mx (with amendment)
- Mx-bx lens composition: Building sync systems from components
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## What is Mx and why we need it



- Multiview modelling in DB and SE
- One UML model = M view models
- Safety assurance: a set /megamodel D of design models, a set $\mathbf{S}$ of safety models, and a set of $\mathbf{N}$ of normative docs are all inter-related and to be kept in sync (megamodel sync).
- Some (many?) binary sync case are actually m-ary



## Change propagation with amendment

- Change propagation to restore consistency may involve amendment of the initial update
- due, mainly, to different granularity of updates in different model spaces
$-\boldsymbol{u}_{\boldsymbol{I}}{ }^{\left({ }^{( }\right)}$is to be consistent with $\boldsymbol{u}_{\boldsymbol{I}}$ : no undoing



## An $m x$ lens with amendment comprises the following data:

- N spaces of models and updates (determined by metamodels, which are implicit))
- Corrs between spaces
- Propagation operations, e.g.,
- $\mathrm{ppg}_{12}$ : Corr x $\mathrm{U}_{1}-->\mathrm{U}_{2}$,
- $\mathrm{ppg}_{13}$ : Corr x $\mathrm{U}_{1}-->\mathrm{U}_{3}$,
- $\mathrm{ppg}_{11}$ : Corr x $\mathrm{U}_{1}$--> $\mathrm{U}_{1}$ (amend.)
- $\mathrm{ppg}_{1^{*}}$ : Corr x $\mathrm{U}_{1}$--> Corr

|  | source |  |  |
| :---: | :---: | :---: | :---: |
| $\stackrel{+}{ \pm}$ | 11 | 21 | 31 |
| $\stackrel{\circ}{\infty}$ | 12 | 22 | 32 |
| $\stackrel{\oplus}{+}$ | 13 | 23 | 33 |

- (with a lot of incidence conditions)
- A lens is called well-behaved (wb) if it satisfies a set of equational laws generalizing those for the binary case


## From state- to delta-based mx with $\mathrm{m} \mathrm{x}_{\mathrm{D}}$ lenses



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## Why we need lens composition: Example



## Why we need lens composition: Example

## World

| Data |
| :---: |
| source 1 |



Formal model: Model spaces (nodes) and lenses (arrows)

Inside lenses: How operations over model states \& updates are composed


## Why we need lens composition: Example

## World



Formal model:
Model spaces (nodes) and lenses (arrows)


Inside lenses: Hiding details


## Why we need lens composition: Example

## World



Formal model:
Model spaces (nodes) and lenses (arrows)


Theorem (composing binary mx).

- Composition $M X=b x 1 \circ m x \circ b x 2$ is an $m x$-lens.
- If some (junction) conditions are satisfied, $M X$ is well-behaved as soon as lenses bx1, mx12, and bx2 are such


## Lens composition types



See: Diskin, König, Lawford: Multiple Model Synchronization with Multiary Delta Lenses with Amendment and K-Putput @ FAOC(2019)

This talk

## Complex sync of multiple systems



## (co)Star composition of lenses



## Theorem (coStar)

- Composition MX = mx* $\left(b x_{1}+\ldots+b x_{n}\right)$ is an mx -lens.
- If some (junction) conditions are satisfied, $\mathbf{M X}$ is well-behaved as soon as lenses $\mathbf{m x}$ and $\mathbf{b x} \mathbf{x}_{\mathbf{i}}$ are such

Theorem (Star—reverse all bx)

- Composition MX is an mx-lens. But well-behavedness is only weakly preserved (see the paper for details).


## Complex sync of multiple systems



## Category theory of lens composition

- Bx Workshops series Bx'12-21
- Michael Johnson, Robert D. Rosebrugh:
- Multicategories of Multiary Lenses. Bx@PLW 2019: 30-44
- Cospans and symmetric lenses. Programming 2018;
- Spans of Delta Lenses. Bx@STAF 2015: 1-15
- Applied Category Theory conference series, ACT'19-21
- Bryce Clark: A diagrammatic approach to symmetric lenses @ ACT'20
- Emma Chollet, Bryce Clarke, Michael Johnson, Maurine Songa, Vincent Wang, Gioele Zardini: Limits and colimits in a category of lenses, ACT'21
- Sophie Libkind, Andrew Baas, Evan Patterson, James Fairbanks, Operadic Modeling of Dynamical Systems: Mathematics and Computation @ ACT'21 (state-based lenses are employed)
- The result on lens composition presented above are from [1], [2], and Multiple Model Synchronization .... @ FAOC(2019) by Zinovy Diskin, Harald König, Mark Lawford


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## Concurrent updates

|  | source |  |  |
| :---: | :---: | :---: | :---: |
| $\begin{aligned} & \stackrel{\rightharpoonup}{ \pm} \\ & \stackrel{\rightharpoonup}{0} \\ & \underset{\sim}{0} \end{aligned}$ | $\mathrm{ppg}_{11}$ | $\mathrm{ppg}_{21}$ | $\mathrm{ppg}_{31}$ |
|  | $\mathrm{ppg}_{12}$ | $\mathrm{ppg}_{22}$ | $\mathrm{ppg}_{32}$ |
|  | $\mathrm{ppg}_{13}$ | $\mathrm{ppg}_{23}$ | $\mathrm{ppg}_{33}$ |

- Possible conflicts between:
- $\operatorname{ppg}_{13}(\mathrm{u} 1)$ and u3
- u1 and $\mathrm{ppg}_{31}(\mathrm{u} 3)$
$-\operatorname{ppg}_{12}(\mathrm{u} 1)$ and $\mathrm{ppg}_{32}(\mathrm{u} 3)$
- Sync needs:
- Conflict resolution
- Update merging


## Approached in:

- Hermann, Ehrig, Ermel, Orejas @ FASE'12
- Xiong, Song, Hu, Takeichi @ SoSym (2013)
- Trollmann \& Albayrak (MGG) @ ICMT'15-17
- Orejas, Pino, Navarro @ FASE'20
- Fritsche, Kosiol, Möller, Schürr, Taentzer @SLE'20
- Future work: employ mx lenses with amendment

Concurrency again: Lens composition with cycling


## Building MX needs conflict resolution and update merging

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## Delta Lenses \& Deltas have been widely used

TGG-based


Model Synch [MODELS11, FASE12, ICGT12, ICMT14, JOT21, ...]


| smm | Buma | PHP |
| :---: | :---: | :---: |
| come | mitmem | Page |
| \%mumper | mmmame | Change |
|  |  | [FSE12] |



API
Updating [ECOOP16]
Incremental Parsing and Tokenizing [MODELS14]



Database View Update [ICFP18]

Business
Process
Model
Synch
[Sosym13]

[^0]45

## Application Example: Program Repair

- Traditional neural program repair approaches use state-based representation of changes

cfa.createEdge(fromNode, Branch.UNCOND, finallyNode);

1. Need to learn delta discovery during training 2. Need to generate 13 tokens correctly
cfa.createEdge(fromNode, Branch.ON_EX, finallyNode);

## Application Example: Program Repair

- Delta-based change representation is simpler and more concise



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- Reverse gradient descent as a lens
- Deep learning and lens composition
- Open games, cybernetics


## Background: String/block diagram notation for functions



## Reverse gradient descent as a lens

(from Cruttwell, Gavranović, Ghani, Wilson, Zanasi: Categorical Foundations of Gradient-Based Learning @ ACT’21)


Asymmetric lens
(reverse gradient descent as put)


Parametric asymmetric lens (reverse gradient as (put ${ }_{A}$, put $_{p}$ )

## ML via parametric lens composition

(from Cruttwell, Gavranović, Ghani, Wilson, Zanasi: Categorical Foundations of Gradient-Based Learning @ ACT’21)


Fig. 3. Model composed with a loss function and a learning rate

## Deep learning via parametric lens composition

- Brendan Fong, David Spivak, Rémy Tuyéras: Backprop as Functor: A compositional perspective on supervised learning @LICS 2019
- Brendan Fong, Michael Johnson: Lenses and Learners @Bx/PLW 2019
- Zinovy Diskin: General Supervised Learning as Change Propagation with Delta Lenses @ FoSSaCS 2020
- David Spivak: Learners' languages @ ACT'21


## Other applications of lenses in AI

- Open games via lenses (state-based)
- Elena Di Lavore, Jules Hedges and Paweł Sobociński Compositional modelling of network games @ Computer Science Logic 2021
- Neil Ghani, Jules Hedges, Viktor Winschel and Philipp Zahn Compositional game theory @ LICS'2018
- Joe Bolt, Jules Hedges, Philipp Zahn: Bayesian open games @arxiv'19
- Categorical cybernetics via lenses (state-based)
- Matteo Capucci, Bruno Gavranovic, Jules Hedges, Eigil Rischel, "Towards foundations of categorical cybernetics" @ ACT'21
- From state- to delta-based lenses in Al (work in progress :-)


## What to take home

- State-based sync is a composed construct: it is a composition of delta discovery and delta propagation. To manage it better, let's separate concerns
- Delta lens composition is a powerful instrument for building complex model synchronizers from components
- Delta lenses do have successful applications in DB, SE, PL
- Possibilities for their applications in AI and specifically ML is under investigation, but first results look promising


## 旍 <br> Thank you!

A M M


[^0]:    [1] Diskin, Gholizadeh, Wider, Czarnecki: A three-dimensional taxonomy for bidirectional model synchronization. J. Syst. Softw. 111: 298-322 (2016)

