Sketch Theory as a Framework for Knowledge Management

NASA IV&V Workshop

9-11 September 2014

Morgantown, WV



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## Introduction Sketches Logic Inference Alignment Context on Context on Conclusions Software Conclusions on Context on Context on Conclusions o

- Mathematical Logic
  - Computational complexity of some predicate calculus fragments
  - Complexity of the syntactic category used for knowledge alignment
  - Challenging to develop a human interface
- Databases + SQL
  - Limited notion of context/view (a single table)
  - Static schema
- Semantic Web OWL/RDF + Description Logic
  - Lack of modularity: meta-data, instance data and uncertainty integrated into a monolithic ontology
  - Limited compositional algebra: (disjoint) unions of ontologies
  - Need for constraint-preserving maps
- Sketch Theory
  - Meager computational infrastructure (e.g., relative to Jena)

Introduction ○○●	Sketches 000000	Logic 00	Inference 00000	Alignment 00000	Context 00	Transformatio	ons Software	Conclusio 00	
Sketch	Theor	y Sti	rength	S					
Facets o	of Knov	wledg	e Mode	ls I	Kno	wledge -	Technolog	ies	
Storage Constrai Alignmei Context/ Reasonir Translati	nts nt /Views ng ions	Queri Unce Dyna Softw Decis Huma	ies rtainty mics vare iion-Maki an Interfa	ng ace	<ul> <li>Mathematical Logic (1879)</li> <li>Databases + SQL (1968)</li> <li>Semantic Web OWL/RDF + Description Logic (1999)</li> <li>Sketches (1968/2000) + Q-Sequences (1990)</li> </ul>				
Sketch	n Theo	ry: O	verview		Sket	ch Theo	ry: Streng	ths	
<ul> <li>Mature, graph-based foundation</li> <li>Vertices = classes or relations</li> <li>Edges = type information or maps</li> <li>Constraints/meta-data specified via graph maps (cones/cocones)</li> <li>Sketch maps respect constraints</li> <li>Grew from category theory in 1968</li> <li>Applied to data modeling since 1989</li> </ul>					Visual/g Modular Combina Concise inter-cor Derived Rich cor Dynamic	graphical m rity: data/ atory alget graphical nvertibility concepts v mposable v cs via sket	nodeling concepts/ur ora of sketch inference an with 1st or via CW algo views/conte ch maps	ncertaint nes Id der logic orithm xt	y :

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- A sketch  $(G, \mathcal{D}, \mathcal{L}, \mathcal{C})$  consists of:
  - An underlying graph G
  - A set  $\mathcal{D}$  of diagrams  $B \to G$
  - A set  $\mathcal{L}$  of cones  $L \to G$
  - A set  $\mathcal{C}$  of cocones  $\mathcal{C} \to \mathcal{G}$
- The graph maps express the axioms or semantic constraints.



Introduction Sketches Logic Inference Alignment Context Transformations Software Conclusions of Set-Based Sketch Models

#### Set-based model of a graph

- Each vertex V is mapped to a set M(V).
- Each edge  $V \xrightarrow{e} W$  is mapped to a function  $M(V) \xrightarrow{M(e)} M(W)$ .



Set-based model of a sketch (G, D, L, C)

- A sketch model is, first, a model of the underlying graph G.
- Sketch constraints impose additional requirements on models.
- Expressiveness of the sketch imposes requirements on suitable categories of semantic models.

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• Partial function model of a graph edge Resident  $\xrightarrow{has}$  TribalElement



• Stochastic matrix model of a graph edge Resident  $\stackrel{\text{has}}{\longrightarrow}$  TribalElement



Introduction Sketches Logic Inference Alignment Context Transformations Software Conclusions on Categorical Semantics of Sketches

- Vertices are interpreted as objects
- Edges are interpreted as morphisms
- Classes of constraints (cones and cocones) are distinguished by the shapes of their base graphs.
- Classes of sketches are distinguished by their classes of constraints.
- Like logics and OWL species, these have different expressive powers.

Sketch Class	Set	Partial Func.	Stoch. Matrices	Čencov Cat.	Prob. 0 Refl.	Dempster Shafer	Fuzzy Sets	Convex Sets
Regular	•	•	•	•	•	•	٠	•
Finite Limit	٠	•	×	×	×	×	•	٠
Finite Coproduct	٠	•	•	•	•	٠	•	٠
Entity-Attribute	•	•	×	×	×	×	٠	•
Mixed	٠	•	×	×	×	×	٠	•

#### Small sample of the sketch semantics landscape

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First formulation of civics concepts:

- Two classes: People and Elected officials
- People have Elected representatives via r.
- Elected officials are instances of people via *u*.
- Elected officials represent themselves via a diagram.



• The diagram truncates the infinite list of composites (property chains).  $u \circ r$   $r \circ u$   $u \circ r \circ u$   $r \circ u \circ r$  ...



Alternative formulation of the concepts:

- One class: Citizens
- Citizens have elected representatives via e.
- Elected officials represent themselves via a diagram.



- Number and names of vertices in  $\mathbb{S}_1$  and  $\mathbb{S}_2$  differ.
- The edges *u* and *r* of S<sub>1</sub> have no corresponding edges in S<sub>2</sub>.
- The edge e of S<sub>2</sub> has no corresponding edge in S<sub>1</sub>.



- Sorts: People, Elected
- Function symbols:
  - $u: \mathsf{Elected} \longrightarrow \mathsf{People}$  r

 $r: \mathsf{People} \longrightarrow \mathsf{Elected}$ 

• Axiom: elected officials represent themselves

 $\top \vdash_x (r(u(x)) = x)$ 





- Sorts: Citizens
- Function symbols:

 $e: \mathsf{Citizens} \longrightarrow \mathsf{Citizens}$ 

• Axiom: elected officials represent themselves

$$\top \vdash_x (e(e(x)) = e(x))$$



#### Set semantics

#### Partial function semantics



#### Logic-Based Inference: Sequent Calculus

Inference

Sketches



Transformations

Software

Conclusions



1	$(u(x) = u(y)) \vdash_{x,y} (u(x) = u(y)) \dots$	Id
2	$(u(x) = u(y)) \vdash_{x,y} \top \dots$	Τ
3	$\top \vdash_{x} (r(u(x)) = x) \dots $	axiom
4	$\top \vdash_{x,y} (r(u(x)) = x) \dots$	Sub (3)
5	$\top \vdash_{x,y} (r(u(y)) = y) \dots $	Sub (3)
6	$(x = y) \land (r(x) = z) \vdash_{x,y,z} (r(y) = z) \dots $	Eq1
0	$(u(x) = u(y)) \land (r(u(x)) = x) \vdash_{x,y,z} (r(u(y)) = x) \dots$	Subs (6)
8	$(u(x) = u(y)) \land (r(u(x)) = x) \vdash_{x,y} (r(u(y)) = x) \ldots \ldots$	Subs (7)
9	$(x = y) \vdash_{x,y} (y = x) \dots$	previous proof
10	$(r(u(y)) = x) \vdash_{x,y} (x = r(u(y))) \dots \dots$	Subs (9)
	$(u(x) = u(y)) \land (r(u(x)) = x) \vdash_{x,y} (x = r(u(y))) \dots \dots \dots$	Cut (8), (10)
12	$(x = y) \land (y = z) \vdash_{x,y,z} (x = z)$	previous proof
13	$(x = r(u(y))) \land (r(u(y)) = y) \vdash_{x,y,z} (x = y) \dots$	Subs (12)
14	$(x = r(u(y))) \land (r(u(y)) = y) \vdash_{x,y} (x = y) \dots \dots \dots \dots$	Subs (13)
15	$(u(x) = u(y)) \vdash_{x,y} (r(u(x)) = x)$	Cut (2), (4)
16	$(u(x) = u(y)) \vdash_{x,y} (u(x) = u(y)) \land (r(u(x)) = x) \ldots \ldots \ldots$	
1	$(u(x) = u(y)) \vdash_{x,y} (x = (r(u(y)))  \dots  \dots  \dots  \dots  \dots  \dots  \dots  \dots  \dots $	Cut (16), (11)
18	$(u(x) = u(y)) \vdash_{x,y} (r(u(y)) = y) \dots $	Cut (2), (5)
19	$(u(x) = u(y)) \vdash_{x,y} (x = r(u(y))) \land (r(u(y)) = y) \ldots \ldots \ldots$	
20	$(u(x) = u(y)) \vdash_{x,y} (x = y) \dots$	Cut (19), (14)

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- Input file: ٩ formulas(assumptions). all x (r(u(x)) = x). end of list. formulas(goals). all x all y  $(u(x) = u(y)) \rightarrow (x = y)$ . end of list. Proof: 1 (all x r(u(x)) = x) .....# label(non clause). [assumption]. 2 (all x all y u(x) = u(y))  $\rightarrow$  x = y .....# label(non clause) # label(goal). [goal]. 3 r(u(x)) = x. .....[clausify(1)]. 4 u(x) = u(y). [deny(2)]. 5 c2 != c1. ......[deny(2)]. 6 x = y. .....[para(4(a,1),3(a,1,1)),rewrite([3(2)])]. 7 \$F. .....[resolve(6,a,5,a)].
- The shorter proof by contradiction uses classical first-order logic.
- First-order horn logic has lower computational complexity in general.

# Introduction Sketches Logic Inference Alignment Context on Sketch-Based Inference: Q-Sequences

- Idea: leverage the notion of *Q*-sequence to implement a reasoning engine for sketch-based knowledge models.
- P. J. Freyd and A. Scedrov. Categories, Allegories. 1990
- D. E. Rydeheard and R. M. Burstall. Computational Category Theory. 1988
- Analogy: Q-sequence proof  $\iff$  logical inference

functional programming  $\iff$  procedural (Haskell) programming (C)

• A *Q*-sequence is a finite list of finitely-presented categories, maps and quantifiers *Q<sub>i</sub>*.



• Satisfaction of a *Q*-sequence in a category (sketch) is defined via a universal mapping property.

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*Q*-Sequence Proof of a Subtype Property in  $\mathbb{S}_1$ 

Alignment

Context

Inference

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In civics sketch  $S_1$ , we may conclude that Elected is a subclass of People.

Software

Conclusions

Transformations

Given any x and y as shown:

Sketches

Introduction

 $Z \xrightarrow{x}_{y} Elected \xrightarrow{u} People$   $Z \xrightarrow{x}_{y} Elected \xrightarrow{u} People \xrightarrow{r} Elected$   $Z \xrightarrow{x}_{y} Elected \xrightarrow{id} Elected$   $Z \xrightarrow{x}_{y} Elected$ 

It follows that u is a monomorphism (one-to-one) in any model.

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Present	ations	;						

- A sketch | first-order theory | ontology is a presentation of knowledge.
- Presentations generate additional knowledge needed for alignment.

Logical theory ${\mathbb T}$	syntactic category $\mathcal{C}_{\mathbb{T}}$
Ontology	rules
$Sketch\ \mathbb{S}$	theory of a sketch $\mathcal{T}(\mathbb{S})$

- Different presentations may generate equivalent structures.
- Sketches S<sub>1</sub> and S<sub>2</sub> representing common concepts are aligned by finding a sketch V and sketch maps as shown.



- Theory of a (linear) sketch
  - Carmody-Walters algorithm for computing left Kan extensions: generalizes Todd-Coxeter procedure used in computational group theory
- Complexity difficult to characterize: can depend on order of constraints www.bakermountain.org/talks/nasa2014.pdf ralphw@bakermountain.org 9–11 September 2014 17/29



$\mathbb{T}_1$	$\mathbb{T}_2$
$u: Elected \to People$	$e: Citizens \rightarrow Citizens$
r: People  o Elected	
$\top \vdash_x (r(u(x)) = x)$	$\top \vdash_x (e(e(x)) = e(x))$



• How can we align the civics theories?

$\mathbb{T}_1$	$\mathbb{T}_2$
u: Elected  o People	$e: Citizens \to Citizens$
r: People  o Elected	
$\top \vdash_x (r(u(x)) = x)$	$\top \vdash_x (e(e(x)) = e(x))$

• Provable equivalence: Every axiom of  $\mathbb{T}_1$  is a theorem of  $\mathbb{T}_2$  and conversely.

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$\mathbb{T}_1$	$\mathbb{T}_2$
u: Elected  o People	$e: Citizens \to Citizens$
r: People  o Elected	
$\top \vdash_x (r(u(x)) = x)$	$\top \vdash_x (e(e(x)) = e(x))$

- Provable equivalence: Every axiom of  $\mathbb{T}_1$  is a theorem of  $\mathbb{T}_2$  and conversely.
- This notion aligns theories that have the same signature.

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$\mathbb{T}_1$	$\mathbb{T}_2$
u: Elected  o People	$e: Citizens \to Citizens$
r: People  o Elected	
$\top \vdash_x (r(u(x)) = x)$	$\top \vdash_x (e(e(x)) = e(x))$

- Provable equivalence: Every axiom of  $\mathbb{T}_1$  is a theorem of  $\mathbb{T}_2$  and conversely.
- This notion aligns theories that have the same signature.
- Alignment typically involves use of derived concepts.

## Introduction Sketches Logic Inference Alignment Context Transformations Software Conclusions on Logic-Based Alignment: Provable Equivalence

$\mathbb{T}_1$	$\mathbb{T}_2$
u: Elected  o People	$e: Citizens \to Citizens$
r: People  o Elected	
$\top \vdash_x (r(u(x)) = x)$	$\top \vdash_x (e(e(x)) = e(x))$

- Provable equivalence: Every axiom of  $\mathbb{T}_1$  is a theorem of  $\mathbb{T}_2$  and conversely.
- This notion aligns theories that have the same signature.
- Alignment typically involves use of derived concepts.
- We need a concept that is less restrictive than provable equivalence.



 Theories T<sub>1</sub> and T<sub>2</sub> are Morita equivalent if their categories of models Mod<sub>T</sub>(D) (in any appropriate semantic category D) are equivalent.

 $\mathsf{Mod}_{\mathbb{T}_1}(\mathcal{D})\,pprox\,\mathsf{Mod}_{\mathbb{T}_2}(\mathcal{D})$ 

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 Theories T<sub>1</sub> and T<sub>2</sub> are Morita equivalent if their categories of models Mod<sub>T</sub>(D) (in any appropriate semantic category D) are equivalent.

$$\mathsf{Mod}_{\mathbb{T}_1}(\mathcal{D})\,pprox\,\mathsf{Mod}_{\mathbb{T}_2}(\mathcal{D})$$

• Theories are Morita equivalent iff their syntactic categories are.

 $\mathcal{C}_{\mathbb{T}_1}\ \approx\ \mathcal{C}_{\mathbb{T}_2}$ 

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 Theories T<sub>1</sub> and T<sub>2</sub> are Morita equivalent if their categories of models Mod<sub>T</sub>(D) (in any appropriate semantic category D) are equivalent.

$$\mathsf{Mod}_{\mathbb{T}_1}(\mathcal{D})\,pprox\,\mathsf{Mod}_{\mathbb{T}_2}(\mathcal{D})$$

- Theories are Morita equivalent iff their syntactic categories are.  $C_{T_1} \approx C_{T_2}$
- This notion solves the alignment problem for our civics theories.

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 Theories T<sub>1</sub> and T<sub>2</sub> are Morita equivalent if their categories of models Mod<sub>T</sub>(D) (in any appropriate semantic category D) are equivalent.

$$\mathsf{Mod}_{\mathbb{T}_1}(\mathcal{D})\,pprox\,\mathsf{Mod}_{\mathbb{T}_2}(\mathcal{D})$$

• Theories are Morita equivalent iff their syntactic categories are.

 $\mathcal{C}_{\mathbb{T}_1} ~pprox ~\mathcal{C}_{\mathbb{T}_2}$ 

- This notion solves the alignment problem for our civics theories.
- It can be difficult to use in practice: syntactic categories are infinite even for very simple theories.

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 $\bullet$  The syntactic category  $\mathcal{C}_{\mathbb{T}}$  of a theory  $\mathbb{T}$  is constructed as follows:

objects:	$lpha$ -equivalence classes of formulae-in-context: $\{ec{x}.arphi\}$				
morphisms :	$\{\vec{x}.\varphi\} \xrightarrow{[\theta]} \{\vec{y}.\psi\}$				
	$\theta \vdash_{\vec{x}, \vec{y}} \varphi \land \psi \qquad \varphi \vdash_{\vec{x}} (\exists \vec{y}) \theta \qquad \theta \land \theta[\vec{z}/\vec{y}] \vdash_{\vec{x}, \vec{y}, \vec{z}} (\vec{z} = \vec{y})$				
composition:	$\{\vec{x}.\varphi\} \xrightarrow{[\theta]} \{\vec{y}.\psi\}$				
	$[(\exists \vec{y})(\theta \land \gamma)] \qquad \qquad [\gamma]$				
	$\{\vec{z}.\chi\}$				
identity:	$\{\vec{x}.\varphi\} \xrightarrow{[\varphi \land (\vec{x}'=\vec{x})]} \{\vec{x'}.\varphi[\vec{x'}/\vec{x}]\}$				

• We restrict the formulae  $\varphi$  and  $\theta$  to be of the appropriate class: cartesian/regular/coherent/first-order.



#### Alignment of the Civics Sketches





• A view  $\mathcal{V} \Longrightarrow \mathbb{S}$  of a sketch  $\mathbb{S}$  is a sketch  $\mathcal{V}$  and a sketch map



- A model of S induces a model of  $\mathcal{T}(S)$  and of its views  $\mathcal{V} \to \mathcal{T}(S) \xrightarrow{M} \text{Set}$ .
- Views may be composed  $\mathcal{V}_2 \Longrightarrow \mathcal{V}_1 \Longrightarrow \mathbb{S}$ .
- View Update Problem: Under what conditions can updates to a model of  $\mathcal{V}$  be propagated to a model of  $\mathbb{S}$  ?

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# Introduction Sketches Logic Inference Alignment Context 00 Context 00 Context 00 Context 00 Context 00 Context 00 Contexts

- Research area with narrower scope: context-sensitive Internet search
  - Google patent for "methods, systems and apparatus including computer program products, in which context can be used to rank search results" (USPTO 8,209,331 2012)
  - Yandex personalized web search challenge: www.kaggle.com
- Techniques to infer context from activities and rank data elements
  - Variable-length hidden Markov model
  - Parametric models of users
  - RankNet, LambdaRank, RankSVM
- Performance metrics used for context-sensitive rankings
  - Normalized discounted cumulative gain (scoring in Kaggle competition)
  - Kendall's au comparison of rankings
  - Jaccard distance between top N rankings and target

Sketch Maps and Model Maps

Sketches

• A sketch map  $\mathbb{S}_1 \to \mathbb{S}_2$  is a graph map  $G_1 \longrightarrow G_2$ 

Inference

that preserves all the constraints of  $\mathbb{S}_1$ .  $B \longrightarrow G_1 \longrightarrow G_2$ 

- We use sketch maps to formulate the alignment problem.
- Given models  $M_1$  and  $M_2$  of a sketch  $\mathbb{S}$ , a model map  $M_1 \to M_2$  is a collection of morphisms (one for each vertex V of G)

Alignment

$$M_1(V) \xrightarrow{\tau_v} M_2(V)$$

that are consistent with the edges of G.

• Example:

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Transformations

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Software

Conclusions

Transfo	rming	Sket	ches in	nto Log	ical T	heories		
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- Sketches are related to first-order logical theories by theorems of the form: Given any sketch S of class X, there is a logical theory T of class Y for which S and T have equivalent classes of models.
- D2.2 of Johnstone's *Sketches of an Elephant: A Topos Theory Compendium* gives explicit constructions of  $\mathbb{T}$  from  $\mathbb{S}$  and conversely.

Class of	Fragment of	
Sketches	Predicate Calculus	Logical Connectives
finite limit	cartesian	$=, \top, \land, \exists^*$
regular	regular	=, ⊤, ∧, ∃
coherent	coherent	=, $\top$ , $\land$ , $\exists$ , $\bot$ , $\lor$
geometric	geometric	=, $\top$ , $\land$ , $\exists$ , $\bot$ , $\lor$
		$\infty$
$\sigma\text{-coherent}$	$\sigma ext{-coherent}$	=, $ op$ , $\wedge$ , $\exists$ , $\perp$ , $\bigvee$
finitary	$\sigma ext{-coherent}$	<i>i</i> =1

\* In cartesian logic, only certain existentially quantified formulae are allowed.

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 Set-based models of entity-attribute sketches can be implemented using database features

Sketch	database schema
Vertex	table with automatically-generated (Serial) key
$Edge\; A \stackrel{e}{\longrightarrow} B$	foreign key in A-table referencing B-table key
Constraints	triggers

- Challenge: manage distributed sketch models, views and constraints Google Megastore, Tenzing and Spanner; Apache Cassandra and Accumulo
- Reasoning
  - Transform to first-order theory then employ theorem prover
  - Q-sequence reasoning using computation category theory tools: Rydeheard and Burstall (ML implementations) 1988
- Theory of a (linear) sketch
  - Carmody-Walters algorithm for computing left Kan extensions: generalizes Todd-Coxeter procedure used in computational group theory
  - Complexity difficult to characterize: can depend on order of constraints

#### 

#### Easik Tool for Modeling with Sketches

- Entity Attribute Sketch Implementation Kit (Easik)
- http://mathcs.mta.ca/research/rosebrugh/Easik
- Build sketches, views and constraints
- Interface with MySQL or Postgres for (set-based models)
- No reasoning engine



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Prograi	n							

- The sketch data model demonstrates valuable features
  - Functional paradigm: syntax, models and maps
  - Separation of meta-data from instance data
  - Uncertainty and lack of information accounted for in models
  - Context-sensitive views which can be composed and combined
  - Formulation of the alignment problem using a well-defined mathematical construction (theory of a sketch)
  - Reasoning via graphical Q-sequences or transformation to predicate calculus fragment
- Research challenges
  - Implement sketch constraints on large, distributed models
  - Leverage insights, datasets and performance metrics from the narrower problem of context-sensitive Internet search
  - Develop and implement semi-automated alignment tool
  - Integrate reasoning and modeling algorithms with instance data into a common software platform
  - Characterize sketch classes corresponding to OWL species

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Referen	ices							

- M. Barr and C. Wells. Toposes Triples and Theories. Springer-Verlag. 1985
- M. Barr and C. Wells. Category Theory for Computing Sciences. Prentice-Hall. 1990
- S. Carmody, M. Leeming and R. F. C. Walters. The Todd-Coxeter Procedure and Left Kan Extensions. J. Symbolic Computation. 19:459–488. 1995
- P. Freyd and A. Scedrov. *Categories Allegories*. North-Holland. 1990
- J. W. Gray. The Category of Sketches as a Model for Algebraic Semantics. In *Categories in Computer Science and Logic*. V. 92 of Contemporary Mathematics. AMS. 1989
- M. Johnson and R. Rosebrugh. Sketch Data Models, Relational Schema and Data Specifications. Electronic Notes in Theoretical Computer Science. 61(6):1–13. 2002
- M. Johnson, R. Rosebrugh and R. J. Wood. Lenses, Fibrations and Universal Translations. Mathematical Structures in Computer Science. 22:25–42. 2012
- P. E. Johnstone. *Sketches of an Elephant: A Topos Theory Compendium*. Oxford University Press. 2002
- F. W. Lawvere and S. Schanuel. *Conceptual Mathematics*. Cambridge University Press. 2<sup>nd</sup> Ed. 2009
- S. Mac Lane. Categories for the Working Mathematician. 2<sup>nd</sup> Ed. Springer-Verlag. 1999
- O. E. Rydeheard and R. M. Burstall. Computational Category Theory. Prentice-Hall. 1988

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