The Geometry of Computation-Graph Abstraction

<u>Koko Muroya</u> (University of Birmingham & RIMS, Kyoto University) Steven W. T. Cheung Dan R. Ghica (University of Birmingham)

LICS 2018 (Oxford), 12 July 2018

an EDSL, for machine learning, with computation graphs

Idealised TensorFlow

Background: TensorFlow (<u>https://www.tensorflow.org/</u>)



Background: TensorFlow (<u>https://www.tensorflow.org/</u>)



Background: Computation in TensorFlow

an EDSL, for machine learning, with computation graphs



example: simple linear regression f(x) = W * x + b

Background: Computation in TensorFlow



Background: Computation in TensorFlow



Background: Programming in TensorFlow



- ✓ automatic collection of cells
- ✓ automatic differentiation
- limited integration with a host language
- imperative update of cells

an EDSL, for machine learning, with computation graphs

- ✓ automatic collection of cells
- ✓ automatic differentiation
 - limited integration with a host language
- imperative update of cells

Idealised TensorFlow

- automatic collection of cells
- automatic differentiation
- ✓ full integration
- *functional* update of cells









Graph abstraction



Muroya (U. B'ham. & RIMS, Kyoto U.)

Graph abstraction



Functional, not imperative, training



Functional, not imperative, training





• built a model using cells

let $(model', p) = abs (\lambda x.\{1\} \times x + \{0\})$ let y = model' (optimiser data p model' loss_function) 7

- built a model using cells
- get a *decoupled* model using graph abstraction

let (model', p) = abs
$$(\lambda x.\{1\} \times x + \{0\})$$

let y = model' (optimiser data p model' loss_function) 7

- built a model using cells
- get a *decoupled* model using graph abstraction
- train a model, by calculating new values of cells

let (model', p) = abs
$$(\lambda x.\{1\} \times x + \{0\})$$

let y = model' (optimiser data p model' loss_function) 7

- built a model using cells
- get a *decoupled* model using graph abstraction
- train a model, by calculating new values of cells
- use the trained model, yielding output

let (model', p) = abs
$$(\lambda x.\{1\} \times x + \{0\})$$

let y = model' (optimiser data p model' loss_function) 7

simply-typed λ -calculus, extended by:

- cells ... containing field elements
- graph abstraction `Abs` ... in implication form `A`
- opaque vectors ... with operations



Proposed Semantics of Idealised TensorFlow

a "dynamic Geometry of Interaction machine" [M. & Ghica '17], a token-guided graph-rewriting abstract machine

- *redex searching* by moving the "token"
- *rewriting* by replacing a sub-graph
- observing value by looking at data carried by the "token"

implementing call-by-value evaluation,

Proposed Semantics of Idealised TensorFlow

a "dynamic Geometry of Interaction machine" [M. & Ghica '17], a token-guided graph-rewriting abstract machine

- redex sear
- rewriting
- observing "token"

computation with computation graphs

implementing call-by-value evaluation,

Proposed Semantics of Idealised TensorFlow

a "dynamic Geometry of Interaction machine" [M. & Ghica '17], a token-guided graph-rewriting abstract machine

- redex sear
- rewriting
- observing "token"

computation with computation graphs

implementing call-by-value evaluation,

... to prove

- type-soundness
- program equivalence
 - restricted call-by-value beta-law
 - garbage collection

an EDSL, for machine learning, with computation graphs

- ✓ automatic collection of cells
- ✓ automatic differentiation
 - limited integration with a host language
- imperative update of cells

Idealised TensorFlow

- automatic collection of cells
- automatic differentiation
- ✓ full integration
- *functional* update of cells

an EDSL, for machine learning, with computation graphs



- autom
- limited
- imper

online visualiser https://cwtsteven.github.io/GoI-TF-Visualiser/C

- BV-with-CBN-embedding/index.html
- ... to see Idealised Tensorflow in full action!

Idealised TensorFlow

- automatic collection of cells
- automatic differentiation
- ✓ full integration
- *functional* update of cells

an EDSL, for machine learning, with computation graphs



Idealised TensorFlow

- automatic collection of cells
- automatic differentiation
- ✓ full integration
- *functional* update of cells

Conclusion

calculi with higher-order computation graphs?

token-guided graph rewriting ("dynamic Geometry of Interaction), powerful & flexible operational semantics

✓ automatic cell collection in TensorFlow

language support for the bureaucracy of model parameters in machine learning

- automatic differentiation in TensorFlow
- dependency between cells à la self-adjusting computation