Good afternoon, my name is Daniil and I'm going to present our joint work with professor Neil Deaton Jones, called "Partial Evaluation and Normalization by Traversals".

=== slide ??: introduction =============

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#TODO: Reformulate: game semantics for various versions of lambdas.

#The game semantics for PCF can be thought of as a PCF interpreter. #In a set of game semantics papers the denotation of an expression #is a game strategy. When played, the game results in a traversal.

For example, Luke Ong's recent paper "Normalization by Traversals" shows that a  $simply-typed\ lambda-expression\ M$  can be evaluated (i.e. normalised) by the algorithm that constructs a traversal of M.

A traversal in this case is a justified sequence of subexpressions of M. For brivety, we will call them tokens. One can think, that M is a program and token is a program point. Any token may have a back pointer, or justifier, to some other previously encountered token. These pointers are used to lookup some information about  $\lambda$ -expressionin the history of computation and to find dynamic binders for variables.

Hereafter we will call this approach to normalization of simply typed lambda terms Oxford Normalization Procedure, or ONP. For a given  $\lambda$ -expression M ONP constructs a set of traversals  $\mathfrak{Trav}(M)$ , each of which corresponds to some  $\underline{\text{path}}$  in normalized term tree

# Confirm the understanding of operational view and its motivation. In the context of our research, we are interested in *operational*, algorithmic view of game semantics-based normalization without direct reference to game semantic foundations (?).

Consider a traversal tr from  $\mathfrak{Trav}(M)$ , which is a sequence of tokens  $t_0$   $t_n$ .

For all such traversals from  $\mathfrak{Trav}(M)$  *ONP* adds on each step zero or more extensions to  $\mathfrak{Trav}(M)$ . If none is added, then the traversal tr already represents some path in the tree of  $\beta$ -normal  $\eta$ -long form of the term.

Each extension is a traversal, obtained by concatenation of traversal tr and <u>one new token</u> t', which may be equipped with a pointer to some previous token in tr.

Moreover, *ONP* uses inference rules, discriminated solely by sintax of end-tokens  $t_n$ . In other words, *ONP* is syntax-directed.

In terms of implementation, we have two main  $\frac{\text{data types}}{\text{a pair}}$ : traversals and items.  $\frac{\text{Traversal}}{\text{backpointer}}$  is just a list of  $\frac{\text{items}}{\text{items}}$ , where an item is  $\frac{\text{a pair}}{\text{a pair}}$ : a token and its (optional) backpointer.

==== slide ??: SOME ONP CHARACTERISTICS ===========

Now then, let's summarize some importants for us properties of ONP.

First, *ONP* can be applied only to simply-tiped  $\lambda$ -terms in  $\eta$ -long form. The  $\eta$ -long form is obtained by  $\eta$ - expanding term fully and replacing implicit binary application operator of each redex by the *long application operator* @. The crucial point is that in order to perform this expansion one needs to know the exact types of all subterms.

Second, *ONP* implements <u>complete</u> head linear reduction. Complete head linear reduction can be seen as regular *head* linear reduction followed by regular *head* linear reductions of all arguments.

The correctness of normalization by traversals is proven in terms of game semantics and categories, and fully based on M's types.

Then, in contrast to standart evaluation approaches, normalization by traversals uses  $\underline{no} \beta$ -reductions, leaves the original term intact, and can be implemented without resorting to *traditional techniques* like environments, closures etc.

Finally, an important property of *ONP* is that while running, *ONP* does not use type information at all. As it was mentioned before, *ONP* constructs traversals using only syntax-derected rules.

And now we are able to formulate the **goals** of our research:

- The first goal is to extend ONP to the untyped case (hereafter UNP);
- The second is to reexamine the outcomes of partial evaluation in the light of alternative evaluation technique.

Now let breafly remined what is partial evaluation and disscuss an effect of its application to normalizer.

Partial evaluation is a program optimization technique also known as program specialization. A one-argument function can be obtained from one with two arguments by specialization, i.e. by "freezing" one input to a fixed value. A partial evaluator by given subject program together with part of its input data s constructs a new program  $p_s$  which, when given p's remaining input d, will yield the same result that p would have produced given both inputs.

- Partial evaluation yeilds program speed up by precomputing all static input at compile time.
- Thus, while talking about partial evaluation data s is called static, data d dynamic, and program  $p_s$  that is build by specializer is called residual.
- A net effect of partial evaluation is a staging program transformation. Partial evaluation devides one-stage program computation in two stages:
  - 1. optimized residual program generation and
  - 2. running the generated program on some dynamic data.
- Most successful applications of PE are compilation and compiler generator. For example, partial evaluation of an interpreter with respect to a source program yields a target program. Moreover, a well-known fact about partial evaluation is that if provided the partial evaluator is self-applicable, then compiler generation is possible by specializing the partial evaluator itself with respect to a fixed interpreter yields a compiler. Finally, specializing the partial evaluator with respect to itself yields a compiler generator.
- An old idea: use partial evaluation to provide Semantics-directed compiler generation. By this we mean more than just a tool to help humans write compilers. Given a specification of a programming language, like a formal semantics or an interpreter, the goal is automatically and correctly transform it into a compiler. The motivation for automatic compiler generation is evident: the three jobs of writing the language specification, writing the compiler, and showing the compiler to be correct are reduced to one: writing the language specification in a form suitable as input to the compiler generator.

The spec equation for a normaliser program  $\bf NP$  that is just a function from lambda-calculus to Traversals can be seen on the slide.

You can see that there is no external dynamic data and that is a tradition for  $\lambda$ -calculus that  $\lambda$ -term M is self-contained.

So a question is **why break normalization into two stages?** Well, ..., there are several reasons for it:

- 1. First, a specialiser output  $NP_M = [spec](NP, M)$  can be in a much simplier language than  $\lambda$ -calculus. And our candidate for it is some **low-level language** called LLL.
- 2. Second, two stages will be natural for *semantics-directed compiler generation*. Our aim is to use *LLL* as an intermediate language to express semantics. Programs on this low-level language can be thought as a semantics for programs from  $\lambda$ -calculus.

In other words, we factor the initial normalization procedure NP into two stages: First stage that called NP $_1$  is a result of partially evaluating normalization procedure to input term M

 $\operatorname{NP}_1 = \llbracket \operatorname{spec} \rrbracket \operatorname{NP} M : \Lambda \to LLL$  and the second stage,  $\operatorname{NP}_2$  is the *semantic function* of LLL-programs  $\operatorname{NP}_2 = \llbracket \cdot \rrbracket : LLL \to Traversals$ .

==== slide ??: how to partially evaluate ONP =======

A next question is "how to partially evaluate oxford nomalization procedure with respect to the static input term M"?

- 1. First, we have to annotate parts of normalization procedure as either static or dynamic. And here variables ranging over
  - (a) tokens that are static, Because there are only finitely many subexpressions of M.

And all other data is actually dynamic

- (b) i.e. back pointers are dynamic;
- (c) and so the traversal being built from both of them is dynamic too.
- 2. Computations in normalization proicedure **NP** are either unfolded by partial evaluator in compile time or residualised that means that partial evaluator will generate a runtime code to do them later in run-time.

  And then
  - Perform all fully static computations at partial evauation time.
  - and for operations to build or test a traversal: generate residual code.

Now we will talk about structure of a spesialized program  $ONP_{M}$ .

Note, that ONP is not quite structually inductive but it is semi-compositional: in a sence that Any recursive ONP call has a substructure of M as argument. This property has several consequences:

- Firts, the partial evaluator can do, at specialisation time, all of the  $O\!N\!P$  operations that depend only on input term M
- wherein this also means that  $ONP_M$  performs no operations at all on lambda expressions(!)

and for all other operations

- the specialised program  $ONP_M$  will be generated. This program contains "residual code", that means that it contains only operations to build the traversal. There are two kind of operation to do this:
  - operations to extend the traversal; and sometimes
  - operations to follow back pointers when needed to do this
- An important fact is that subexpressions of M will appear, but are only used as tokens:

This means that tokens are indivisible: they are only used as labels (i.e. program points) and for equality comparisons with other tokens. Actually we use names instead of real subexpressions. And real subexpressions is only needed for the normalized term reconstruction from traversals.

Statute of our world for sympty supout A carounds to do follows:

- 1. We have one version of *ONP* writen in Haskell and another in Scheme
- 2. The Haskell version includes: typing using algorithm *W*; conversion to eta-long form; the traversals generation algorithm itself; and construction of the normalised term from the set of traversals.
- 3. While Scheme version is nearly ready to apply automatic partial evaluation. We are planning to use the unmix partial evaluator that is written by Sergei Romanenko and others. unmix is a general partial evaluator for Scheme. We expect that we will achieve the described above effect of specialising *ONP*.
- 4. An important fact is that the LLL output program size is only linearly larger than M, satisfying

$$|p_M| = O(|M|)$$

while traversal itself can be unboundaly larger than size of the input term.

5. We have also have a handwritten a *generating extension* of  $ONP \ ONP - gen$ . Symbolically,

If 
$$p_M = [NP-gen]^{scheme}(M)$$
 then  $\forall M . [M]^{\Lambda} = [p_M]^{LLL}$ 

It means that by given  $\lambda$ -term M OMP-gen generates an LLL program  $p_M$  that being executed generates a traversal for M.

For now: *LLL* is a tiny subset of scheme, so the output  $p_M$  is a scheme program.

- 6. There are more thing to do for simply-typed  $\lambda$ -calculus:
  - First, produce a generating extension automatically by specialising the specialiser to a  $\Lambda$ -traverser using unmix.
  - Second, redefine *LLL* formally as a *clean stand-alone tiny first-order subset* of haskell and use haskell supersompiler to achive a partial evaluation effect.
  - Also we whant to extend existing approach to programs with input dynamic data in run-time.

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For untyped  $\lambda$ -calculus

- 1. We have a normaliser *UNP* that is a normaliser for arbitrary untyped lambda term.
- 2. *UNP* has been done in Haskell and works on a variety of examples. Right now we are working on a more abstract definition of *UNP*.
- 3. Some of traversal items may have two back pointers, in comparison: *ONP* uses only one.
- 4. As *ONP*, *UNP* is also defined semi-compositionally by recursion on syntax of  $\lambda$ -expression M. This actually means that it also can be specialized as *ONP* can.
- 5. Moreover, by specialising UNP, an arbitrary untyped  $\lambda$ -expression can be translated to our low-level language. So the specialised version of UNP could be a semantic function for  $\lambda$ -calculus.
- 6. Correctness proof: pending. For now, we are working on a correctness proof for UNP. We expect that we will prove its correctness formally using some proof assistant like **Coq**.

1. An idea is to regard a computation of  $\lambda$ -expression M on input d as a two-player game between the lll-codes for M and d.

- 2. An intresting examples in this case a is usual  $\lambda$ -calculus definition of multiply function (mult) on Church numerals.
- 3. Amaizing fact is that **Loops from out of nowhere**:
  - Neither mult nor the data contain loops;
  - but mult function is compiled into an *LLL*-program with two nested loops, one to each input numeral, that being applied to two Church numerals computes their product.
  - We expect that we can do the computation entirely without back pointers.
- 4. Right now we are trying to express such program-data games in a *communicating* version of *LLL*.