

# Uncomputation

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Program execution need memory. Program may run out of memory for multiple reasons: big dataset, exploding intermediate state, the machine have less memory than others, etc. When this happens, the program either get killed, or the operating system swaps, significantly degrading the performance. We propose a technique, uncomputation, that allow the program to continue running gracefully even after breaching the memory limit, without significant performance degradation. Uncomputation work by turning computed values back into thunk, and upon re-requesting the thunk, computing and storing them back. A naive implementation of uncomputation will face multiple problems. Among them, the most crucial and the most challenging one is that of breadcrumb. After a value is uncomputed, it's memory can be released but some memory, breadcrumb, is needed, so we can recompute the value back. Ironically, in a applicative language, due to boxing all values are small. This mean uncomputation, implemented naively, will only consume more memory, defeating the purpose. We present a runtime system, implemented as a library, that is absolved of the above breadcrumb problem.

Additional Key Words and Phrases: Do, Not, Us, This, Code, Put, the, Correct, Terms, for, Your, Paper

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## 1 INTRO

## 2 OVERVIEW

The tock tree serve as a cache **insert this sentence somewhere**

Section: The CEKR Machine (???) Replay Stack the section with lots of greeks argue about progress

Section: Implementation (3pg long) Heuristic Loop Unrolling

Key question: How to get the replay stack small? Key question: Garbage Collection/Eager Eviction

Summarize the meeting into key step

Double  $O(1)$

## 3 CORE LANGUAGE

Zombie works on a purely functional language with products, sum types, and first-class functions called  $\lambda_Z$ . For simplicity in this paper, we treat the language as untyped. The syntax of  $\lambda_Z$  is shown in Figure 3; its semantics are standard. The full Zombie implementation supports additional features, such as primitive types and input/output; ?? describes how these features are layered on top of the core implementation described here.

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Importantly, because Zombie is purely functional, programs are totally deterministic, in the sense that evaluating a given expression in a given environment always returns the same result. This is essential for Zombie to work correctly.

### 3.1 CEK Machine

The key insight of Zombie is to assign a unique identifier to any value ever allocated during the program's execution. Conceptually, it identifies each value with the execution step that allocated it. It does this using a variant of the CEK machine.

The CEK machine is a well-known abstract machine for executing untyped lambda calculi, where the machine state consists of three parts:

- (1) **C**ontrol, the expression currently being evaluated.
- (2) **E**nvironment, a map from the free variables of the Control to their values.
- (3) **K**ontinuation, which is to be invoked with the value the Control evaluates to.

*We should use  $C$  for expressions, to match the CEK terminology.*

In our variation of the CEK machine, we split the evaluation phase from the invocation of the continuation, resulting in a machine state that looks like this:

$$\text{State} ::= \text{Eval } E \text{ Env } K \mid \text{Apply } K \ V$$

In other words, our CEK machine can be in an Eval or an Apply state; the Eval stores the classic control, environment, and kontinuation, while the Apply state stores just a continuation and a value. The precise syntax of values and kontinuations are given in Figure 4.

Execution in the CEK machine involves a series of transitions between these machine states; in other words, it is a transition system. To run a program  $C$  in the CEK machine, one first sets up an initial state ( $\text{Eval } C \ \{\} \ \text{Done}$ ) whose Control is the expression to evaluate, whose Environment is empty, and whose Kontinuation is a special Done continuation. The rules of the CEK machine are then used to transition from one machine state to another, until finally reaching the state  $\text{Apply Done } V$ . In that state,  $V$  is the result of evaluating of  $C$ .<sup>1</sup>

A notable property of the CEK machine is that each transition performs a bounded amount of work. Contrast this to a traditional operation semantics, where the semantics of a Case statement might be something like:

$$\frac{\Gamma \vdash e \rightarrow^* \text{Left } \Gamma' \vdash V}{\Gamma \vdash \text{Case } e \ x \ e_l \ y \ e_r \rightarrow \Gamma', x : V \vdash e_l}$$

Here the antecedent of the inference rule might perform arbitrarily many steps, and thus an arbitrary amount of computation; derivations thus form a tree. In the CEK machine, no such steps exist, and derivations in the CEK machine form a flat list. This property of the CEK machine is illustrated graphically in Figure 1.

### 3.2 Determinism and Tocks

Importantly, the CEK machine is linear and deterministic. This means that every CEK machine state transitions to at most one state. That in turn means that if we were to “rewind” a CEK machine, putting it in an earlier state, it would transition through the exact same sequence of states in the exact same order. This determinism or “replayability” is essential for Zombie to work, and is illustrated graphically in Figure 2.

<sup>1</sup>Note that, because  $\lambda_Z$  is untyped, it contains non-terminating programs using, for example, the  $Y$  combinator. In the CEK machine, these programs create an infinite sequence of machine states that do not include a terminating  $\text{Apply Done } V$ .

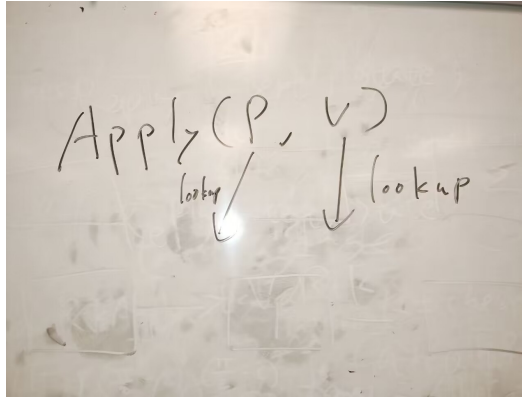


Fig. 1. the machine does a small constant amount of pointer lookup

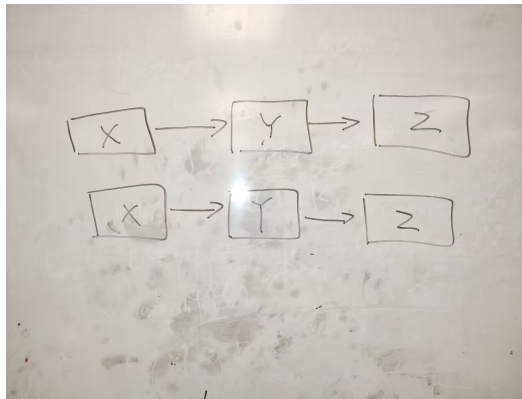


Fig. 2. the deterministic, linear nature of the CEK machine

One key property guaranteed by determinism is that, for a given initial program  $C$ , the machine state at any point during  $C$ 's execution can be uniquely identified by how many steps have been executed since the initial state. In other words, the initial state is identified with the number 0, the next state it transitions to with the number 1, and so on. This state number, which we call a “tock”, is logically unbounded, but in our implementation it is stored as a 64-bit integer. In our implementation, that suffices for several decades of runtime on current hardware.

### 3.3 Heap Memory

Because we are interested the total memory usage of  $\lambda_Z$  program, our variant of the CEK machine includes an explicit heap and explicit pointers. That is, in our formalization values  $V$  are formalized as pointers  $P(\text{VCell})$  to “value cells”. Values cells—which can be closures, products, and sums—in turn contain values, that is, pointers.

During execution, the CEK machine looks up these values using an explicit heap  $H$ . To interact with the heap, the CEK machine uses two functions.  $\text{Lookup} : (P(X), H) \rightarrow X$  dereferences a pointer in the heap.  $\text{Alloc} : (X, H) \rightarrow (P(X), H')$  stores a value in the heap, returning a pointer to it and the updated heap. Transitions in the CEK machine call  $\text{Lookup}$  and  $\text{Alloc}$  in order to make  $\text{Eval}$  and  $\text{Apply}$  steps, as shown in Figure 5 and Figure 6.

Name	$N$	$::=$	A set of distinct names
Expr	$E$	$::=$	$N \mid \text{Let } N E E \mid \text{Lam } N E \mid \text{App } E E \mid$ $\text{Prod } E E \mid \text{Zro } E \mid \text{Fst } E \mid \text{Left } E \mid \text{Right } E \mid$ $\text{Case } E N E N E$

Fig. 3. The syntax of  $\lambda_Z$ .

Continuation	$K$	$::=$	$P(\text{KCell})$
KCell		$::=$	$\text{Done} \mid \text{KLet } N \text{ Env } E K \mid \text{KApp}_0 \text{ Env } E K \mid$ $\text{KApp}_1 \text{ Env } N E K \mid \text{KProd}_0 \text{ Env } E K \mid$ $\text{KProd}_1 V K \mid \text{KZro } K \mid \text{KFst } K \mid \text{KLeft } K \mid$ $\text{KRight } K \mid \text{KCase } \text{Env } N E N E K$
Value	$V$	$::=$	$P(\text{VCell})$
VCell		$::=$	$\text{Clos } \text{Env } N E \mid \text{VProd } V V \mid \text{VLeft } V \mid$ $\text{VRight } V$
Environment	$\text{Env}$	$::=$	$(N, V) \dots$
State		$::=$	$\text{Eval } E \text{ Env } K \mid \text{Apply } K V$

Fig. 4. Definitions for the CEK Machine

Importantly, every CEK machine step (whether Eval or Apply) makes at most two calls to Lookup and Alloc, of which at most one is an Alloc call. This means that each allocation the program makes can be uniquely identified by the tock for the machine state where it is allocated. This identification is the core abstraction that drives Zombie’s implementation.

## 4 UNCOMPUTING AND RECOMPUTING

### 4.1 Tock

The key insight of Zombie is to introduce an abstraction layer between the executing program and the heap. This abstraction layer will allow the heap to transparently discard and recompute the program’s intermediate values. To do so, we use the CEK machine to refer to each intermediate value by the index of the computation step that created it. Since each CEK step allocates at most one cell, this correspondence is injective. In other words, instead of storing pointers that refer to memory locations, we will store pointers, which we call “tokens”, that refer to points in time—that is, CEK step indices.

In our runtime, these “tokens” are implemented as 64-bit integers, though in our model we will treat them as logically unbounded integers. There is a global “current tock” counter, which starts at 0 and is increased by 1 at every transition step in the abstract machine and at every allocation. The cell allocated at step  $i$  is then referred to by the tock  $i$ , and that tock can then be used as a pointer, stored in data structures, looked up by later computations steps, and the like.

Note that, due to the determinism and linearity of the CEK machine, tokens are strictly ordered and the value computed at some tock  $i$  can only depend on values computed at earlier tokens. Moreover we can recreate the value at tock  $i$  by merely rerunning the CEK machine from some earlier state to that point. Because the CEK machine is deterministic, this re-execution will produce the same exact value as the original. Because our pointers refer to abstract values, not to specific memory locations, the heap can thus re-compute a value as necessary instead of storing it in memory.

In other words, the way Zombie works is that the heap will store only some of the intermediate values. The ones that aren’t stored will instead be recomputed as needed, and the heap will store

$$\begin{array}{c}
\frac{}{\text{State}, H \rightsquigarrow \text{State}, H} \qquad \frac{}{\text{Eval}(N, \text{Env}, K), H \rightsquigarrow \text{Apply}(K, \text{Env}(N)), H} \\
\\
\frac{\text{Alloc}(\text{KLeft } K, H) = (P, H')}{\text{Eval}(\text{Left } X, \text{Env}, K), H \rightsquigarrow \text{Eval}(X, \text{Env}, P), H'} \\
\\
\frac{\text{Alloc}(\text{KRight } K, H) = (P, H')}{\text{Eval}(\text{Right } X, \text{Env}, K), H \rightsquigarrow \text{Eval}(X, \text{Env}, P), H'} \\
\\
\frac{\text{Alloc}(\text{KProd}_0 K R, H) = (P, H')}{\text{Eval}(\text{Prod } L R, \text{Env}, K), H \rightsquigarrow \text{Eval}(L, \text{Env}, P), H'} \\
\\
\frac{\text{Alloc}(\text{KZro } K, H) = (P, H')}{\text{Eval}(\text{Zro } X, \text{Env}, K), H \rightsquigarrow \text{Eval}(X, \text{Env}, P), H'} \qquad \frac{\text{Alloc}(\text{KFst } K, H) = (P, H')}{\text{Eval}(\text{Fst } X, \text{Env}, K), H \rightsquigarrow \text{Eval}(X, \text{Env}, P), H'} \\
\\
\frac{\text{Alloc}(\text{KCase } LN L RN R \text{ Env}, H) = (P, H')}{\text{Eval}(\text{Case } X LN L RN R, \text{Env}, K), H \rightsquigarrow \text{Eval}(X, \text{Env}, P), H'} \\
\\
\frac{\text{Alloc}(\text{KLet } A K C \text{ Env}, H) = (P, H')}{\text{Eval}(\text{Let } A B C, \text{Env}, K), H \rightsquigarrow \text{Eval}(B, \text{Env}, P), H'} \qquad \frac{\text{Alloc}(\text{KApp}_0 K X, H) = (P, H')}{\text{Eval}(\text{App } F X, \text{Env}, K), H \rightsquigarrow \text{Eval}(F, P), H'} \\
\\
\frac{\text{Alloc}(\text{Clos } \text{Env}(\text{fv}) \cdots N E, H) = (P, H')}{\text{Eval}(\text{Lam } N E, \text{Env}, K), H \rightsquigarrow \text{Apply}(K, P), H'}
\end{array}$$

Fig. 5. Abstract Machine Transition: Eval

earlier CEK machine states in order to facilitate that. As the program runs, intermediate values can be discarded from the heap to reduce its memory usage.

Because a value can be recomputed from *any* earlier CEK machine state, it will also turn out to be possible to store relatively few machine states. Therefore, overall memory usage—for the stored intermediate values and the stored machine states—can be kept low. In *Zombie*, we can store asymptotically fewer machine states than intermediate values, allowing us to asymptotically reduce memory usage.

## 4.2 Tock Tree

In *Zombie*, the heap is implemented by a runtime data structure called the tock tree. The tock tree maps tocks to the cells allocated by that step; in other words, the tock tree implements the mapping between tocks and their actual memory locations. Because values can be evicted to save memory, however, not all tocks have a mapping in the tock tree. Instead, every tock that *is* present in the tock tree will also store its machine state. To recompute an evicted value at some tock  $i$ , the tock tree will find the largest machine state at tock  $j < i$  and replay from that tock to tock  $i$ .

Each node in the Tock Tree stores both a memory cell and a machine state. Specifically, consider the execution of the CEK machine from step  $t$ . In step  $t + 1$  it allocates a cell; it then performs a computation in step  $t + 2$ . So the node at tock  $t$  contains both the cell allocated at step  $t + 1$  as well as the machine state *afterwards*, at step  $t + 2$ . A transition might not alloc any new cells, in such

$$\begin{array}{c}
\frac{\text{Lookup}(P, H) = \text{KLeft } K \quad \text{Alloc}(\text{VLeft } V, H) = (P', H')}{\text{Apply}(P, V), H \rightsquigarrow \text{Apply}(K, P'), H'} \\
\\
\frac{\text{Lookup}(P, H) = \text{KRight } K \quad \text{Alloc}(\text{VRight } V, H) = (P', H')}{\text{Apply}(P, V), H \rightsquigarrow \text{Apply}(K, P'), H'} \\
\\
\frac{\text{Lookup}(P, H) = \text{KCase } \text{Env } LN \ L \ RN \ R \ K \quad \text{Lookup}(V, H) = \text{VLeft } V}{\text{Apply}(P, V), H \rightsquigarrow \text{Eval}(L, \text{Env}(LN := V), K), H} \\
\\
\frac{\text{Lookup}(P, H) = \text{KCase } \text{Env } LN \ L \ RN \ R \ K \quad \text{Lookup}(V, H) = \text{VRight } V}{\text{Apply}(P, V), H \rightsquigarrow \text{Eval}(R, \text{Env}(RN := V), K), H} \\
\\
\frac{\text{Lookup}(P, H) = \text{KProd}_0 \ \text{Env } R \ K \quad \text{Alloc}(\text{KProd}_1 \ V \ \text{Env } K, H) = (P', H')}{\text{Apply}(P, V), H \rightsquigarrow \text{Apply}(K, P'), H'} \\
\\
\frac{\text{Lookup}(P, H) = \text{KProd}_1 \ L \ K \quad \text{Alloc}(\text{VProd } L \ V, H) = (P, H')}{\text{Apply}(P', V), H \rightsquigarrow \text{Apply}(K, P'), H'} \\
\\
\frac{\text{Lookup}(P, H) = \text{KZro } K \quad \text{Lookup}(V, H) = (\text{VProd } X \ Y)}{\text{Apply}(P, V), H \rightsquigarrow \text{Apply}(K, X), H'} \\
\\
\frac{\text{Lookup}(P, H) = \text{KFst } K \quad \text{Lookup}(V, H) = (\text{VProd } X \ Y)}{\text{Apply}(P, V), H \rightsquigarrow \text{Apply}(K, Y), H'} \\
\\
\frac{\text{Lookup}(P, H) = \text{KLet } A \ \text{Env } C \ K}{\text{Apply}(P, V), H \rightsquigarrow \text{Eval}(C, \text{Env}[A := V], K), H'} \\
\\
\frac{\text{Lookup}(P, H) = \text{KApp}_0 \ \text{Env } X \ K \quad \text{Lookup}(V, H) = (\text{Clos } \text{Env}' \ N \ E, H) \quad \text{Alloc}(\text{KApp}_1 \ \text{Env}' \ N \ E \ K, H)}{\text{Apply}(P, V), H \rightsquigarrow \text{Eval}(X, \text{Env}, P'), H'} \\
\\
\frac{\text{Lookup}(P, H) = \text{KApp}_1 \ \text{Env } N \ E \ K}{\text{Apply}(P, V), H \rightsquigarrow \text{Eval}(E, \text{Env}[N := V], K), H'}
\end{array}$$

Fig. 6. Abstract Machine Transition: Apply

Heap	$H$	::=	An abstract key value store
Pointer $\langle X \rangle$	$P\langle X \rangle$	::=	Key into heap with value type $X$
Lookup	:		$(\text{Pointer}\langle X \rangle, H) \rightarrow X$
Alloc	:		$(X, H) \rightarrow (\text{Pointer}\langle X \rangle, H)$

Fig. 7. Heap API

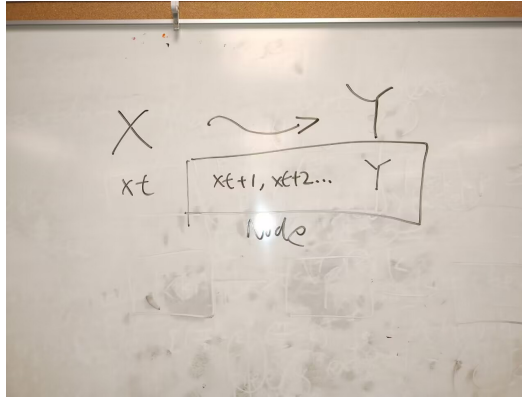


Fig. 8. a node in the tock tree

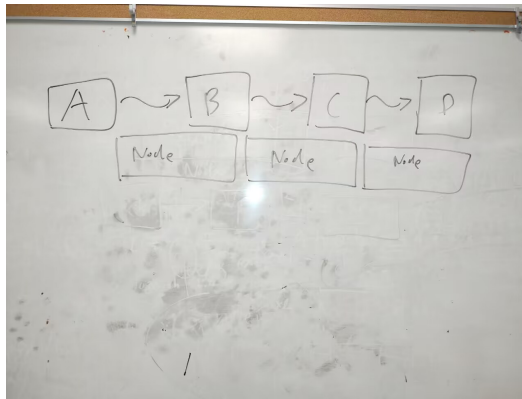


Fig. 9. the tock tree with multiple nodes

a case the Node only store a machine state any no memory cell. To be more specific, if the node represent the execution of the CEK machine at step  $t$ , it will only contain the CEK machine state at step  $t + 1$ .

In order to make this operation efficient, the tock tree is organized as a binary search tree. In a binary search tree, arbitrary keys (tocks) can be looked up in  $O(\log(n))$  time, where  $n$  is the size of the tock tree, and when a key is not found, the largest key smaller than it can be easily found. Then the heap can replay execution from that point to compute the desired tock.

### 4.3 Happy path

Every state transition and pointer lookup is then converted into node insertion and node lookup into this data structure. After we had completed a transition, we construct the node, packing the allocated cells during the transition and the transit-to state, inserting the node onto the tock tree.

Originally, transition might require pointers lookup. Such lookup is translated to a query to the tock tree with the given node. Ideally speaking, the returned node contain the cell that correspond to the given tock. We can then convert the tock into the corresponding cell and continue execution. Note that we still need the tock tree to be lenient in this case, as the key of the node denote the beginning of the transition, not the cell itself.

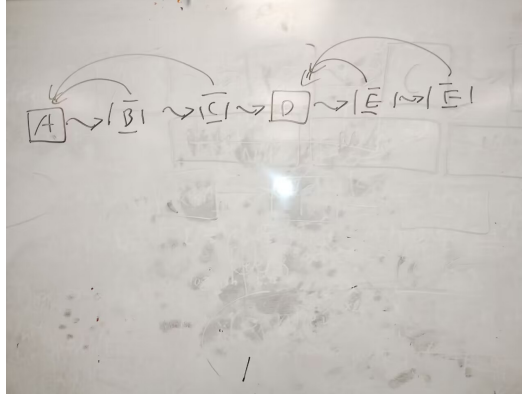


Fig. 10. lookup failure return the latest earlier node

#### 4.4 Sad Path

Sadly, as we had removed nodes from the Tock Tree, we might not be able to retrieve a cell directly, and might need to recompute it - the point of the paper.

To recompute a value at tock  $t$ , the following steps are taken:

- (1) Suspend the current execution into another kind of continuation, Replay Continuation (RK).
- (2) Execute the transit-to state from the looked-up node in the tock tree, until tock reach  $t$ .
- (3) Resume RK with the cell created at tock  $t$ .

Note that the Replay Continuation operate at a more fine-grain granularity then that of the normal continuation. This is because a single transit step might do multiple lookup, but RK need to correspond to a lookup failure in such a transition. Otherwise the requested value might be immediately uncomputed again, and the whole process enter an infinite loop.

After the above 3 steps, the execution shall continue as if no replay had happens at all, and lookup return a node with the cell we wanted. In other words, the happy path and the sad path should converge.

During the replaying process, more lookup might be issued, and those lookup might need more replay - replay is recursive. Just like the classical continuation at the CEK machine, the Replay Continuation need to be recursive, and form a stack as well.

#### 4.5 Reified Continuation

We treat continuations also as values, and label them with tock/put them in the tock tree, just like any other values. This allow us to also uncompute continuation as well.

#### 4.6 Eviction

This allow us to remove any non-leftmost node from the tock tree. After the removal, the query that originally return the removed node, will return the node slightly earlier then that, which we can then replay to regenerate the removed node. In fact, this is the implementation of uncomputation in our system, and any non-leftmost node can be removed, to save memory at any given time.



before	:	Value	=	P<VCell>
after	:	Value	=	Tock
before	:	Continuation	=	P<KCell>
after	:	Continuation	=	Tock
before	:	State	::=	Eval $E \text{ Env } K$   Apply $K \ V$
after	:	State	::=	Eval $E \text{ Env } K \text{ Tock}$   Apply $K \ V \text{ Tock}$
Node			=	(Maybe(KCell   VCell), State)
Query	:	(TockTree, Tock)	->	(Tock, Node)
Insert	:	(TockTree, Tock, Node)	->	TockTree

Fig. 11. Tock Tree API. Note that we deliberately avoid dictating what node get uncomputed, in order to decouple uncomputation/recomputation with selecting what to uncompute. Instead, node might be dropped during insertion into TockTree. One might e.g. set a limit onto the amount of nodes in the Tock Tree.

ReplayContinuation	RK	::=	NoReplay   RKApply $V \text{ Tock } RK$   RKCase $\text{Tock Env } N \ E \ N \ E \ K \ RK$   RKZro $\text{Tock } K \ RK$   RKFst $\text{Tock } K \ RK$   RKApp $\text{Tock Env } E \ K \ RK$
Replay	R	::=	(State, RK)

## 5 CEKR MACHINE

In this section we formalize the semantic of the language with uncomputation and recomputation. It is implemented by adding a Replay Continuation alongside the CEK Machine. Hence we call it the CEKR Machine.

### 5.1 Tock Tree

### 5.2 Replay Continuation

## 6 IMPLEMENTATION

### 6.1 Tock Tree

To exploit the temporal/spatial locality, and the 20-80 law of data access (cite?), the tock tree is implemented as a slight modification of a splay tree.

This design grant frequently-accessed data faster access time. Crucially, consecutive insertion take amortized constant time.

The tock tree is then modified such that each node contain an additional parent and child pointer. The pointers form a list, which maintain an sorted representation of the tock tree. On a query, the tock tree do a binary search to find the innermost node, then follow the parent pointer if that node is greater then the key. This process is not recursive: the parent pointer is guaranteed to have a smaller node then the input key, as binary search will yield either the exact value, or the largest value less then the input, or the smallest value greater then the input.

### 6.2 Picking Uncomputation Candidate

Note that the guarantee we prove is independent of our policy that decide which value to uncompute (eviction policy).

#### 6.2.1 Union Find.

#### 6.2.2 The Policy.

$$\begin{array}{c}
\frac{}{RApply : (Tock, Node, RK) \rightarrow (State, RK)} \quad \frac{}{RApply(\_, (\_, st), NoReplay) = (st, NoReplay)} \\
\\
\frac{}{RApply(\_, (Nothing, st), rk) = (st, rk)} \quad \frac{t + 1! = t'}{RApply((t, Just\ cell, st), rk) = (st, rk)} \\
\\
\frac{t + 1 = t'}{RApply((t, Just\ cell, st), RKApply\ vt'rk) = (Apply\ vcellt', rk)} \\
\\
\frac{t + 1 = t'}{RApply((t, Just\ VLeftX, st), RKCase\ t'\ Env\ LN\ L\ RN\ R\ K\ RK) = ((Eval\ L\ Env[LN := X]K, t'), rk)} \\
\\
\frac{t + 1 = t'}{RApply((t, Just\ VRightX, st), RKCase\ t'\ Env\ LN\ L\ RN\ R\ K\ RK) = (Eval\ R\ Env[RN := Y]Kt', rk)} \\
\\
\frac{t + 1 = t'}{RApply((t, Just\ VProd\ X\ Y, st), RKZro\ t'\ K\ RK) = (Apply\ X\ K\ t', RK)} \\
\\
\frac{t + 1 = t'}{RApply((t, Just\ VProd\ X\ Y, st), RKfst\ t'\ K\ RK) = (Apply\ Y\ K\ t', RK)} \\
\\
\frac{t + 1 = t'}{RApply((t, Just\ ClosEnv'\ N\ E, st), RKApp\ t'\ Env\ X\ K\ RK) = (EvalX\ Env\ (KApp1Env'\ NEK), RK)}
\end{array}$$

### 6.2.3 GDSF.

## 6.3 Language Implementation

For implementation simplicity and interoperability with other programs, zombie is implemented as a C++ library, and the Cells are ref-counted. Our evaluation compiles the program from the applicative programming language formalized above(give name), to C++ code.

## 6.4 Optimization

**6.4.1 Fast access path.** Querying the tock tree for every value is slow, as it requires multiple pointer traversal. To combat this, each Value is a Tock paired with a weak reference, serving as a cache, to the Cell. When reading the value, if the weak reference is ok, the value is return immediately. Otherwise the default path is executed, and the weak reference is updated to point to the new Result.

**6.4.2 Loop Unrolling.** To avoid frequent creation of node object, and their insertion to the tock tree, multiple state transition is packed into one.

## 6.5 Bit counting

# 7 FORMAL GUARANTEE

## 7.1 Safety

Evaluating under replay semantic give same result as under normal semantic

$$\begin{array}{c}
\hline
(\text{State}, rk), \text{TockTree} \rightsquigarrow (\text{State}, rk), \text{TockTree} \\
\hline
\frac{st = \text{Apply}(K, \text{Env}(N), t + 1) \quad node = (\text{Nothing}, st) \quad TT' = \text{insert}(TT, t, node)}{(\text{Eval}(N, \text{Env}, K, t), rk), TT \rightsquigarrow \text{RApply}(t, node, rk), TT')} \\
\hline
\frac{st = \text{Eval}(X, \text{Env}, t + 1, t + 2) \quad node = (\text{Just KLeft } K, st) \quad TT' = \text{insert}(TT, t, node)}{(\text{Eval}(\text{Left } X, \text{Env}, K, t), rk), TT \rightsquigarrow \text{RApply}(t, node, rk), TT')} \\
\hline
\frac{st = \text{Eval}(X, \text{Env}, t + 1, t + 2) \quad node = (\text{Just KRight } K, st) \quad TT' = \text{insert}(TT, t, node)}{(\text{Eval}(\text{Right } X, \text{Env}, K, t), rk), TT \rightsquigarrow \text{RApply}(t, node, rk), TT')} \\
\hline
\frac{st = \text{Eval}(X, \text{Env}, t + 1, t + 2) \quad node = (\text{Just KProd}_0 K R, st) \quad TT' = \text{insert}(TT, t, node)}{(\text{Eval}(\text{Prod } L R, \text{Env}, K, t), rk), TT \rightsquigarrow \text{RApply}(t, node, rk), TT')} \\
\hline
\frac{st = \text{Eval}(X, \text{Env}, t + 1, t + 2) \quad node = (\text{Just KZro } K, st) \quad TT' = \text{insert}(TT, t, node)}{(\text{Eval}(\text{Zro } X, \text{Env}, K, t), rk), TT \rightsquigarrow \text{RApply}(t, node, rk), TT')} \\
\hline
\frac{st = \text{Eval}(X, \text{Env}, t + 1, t + 2) \quad node = (\text{Just KFst } K, st) \quad TT' = \text{insert}(TT, t, node)}{(\text{Eval}(\text{Fst } X, \text{Env}, K, t), rk), TT \rightsquigarrow \text{RApply}(t, node, rk), TT')} \\
\hline
\frac{st = \text{Eval}(X, \text{Env}, t + 1, t + 2) \quad node = (\text{Just KCase } LN L RN R \text{ Env}, st) \quad TT' = \text{insert}(TT, t, node)}{(\text{Eval}(\text{Case } X LN L RN R, \text{Env}, K, t), rk), TT \rightsquigarrow \text{RApply}(t, node, rk), TT')} \\
\hline
\frac{st = \text{Eval}(X, \text{Env}, t + 1, t + 2) \quad node = (\text{Just KLet } A K C \text{ Env}, st) \quad TT' = \text{insert}(TT, t, node)}{(\text{Eval}(\text{Let } A B C, \text{Env}, K, t), rk), TT \rightsquigarrow \text{RApply}(t, node, rk), TT')} \\
\hline
\frac{st = \text{Eval}(X, \text{Env}, t + 1, t + 2) \quad node = (\text{Just KApp}_0 K X, st) \quad TT' = \text{insert}(TT, t, node)}{(\text{Eval}(\text{App } F X, \text{Env}, K, t), rk), TT \rightsquigarrow \text{RApply}(t, node, rk), TT')} \\
\hline
\frac{st = \text{Eval}(X, \text{Env}, t + 1, t + 2) \quad node = (\text{Just Clos } \text{Env}(\text{fv}) \cdots N E, st) \quad TT' = \text{insert}(TT, t, node)}{(\text{Eval}(\text{Lam } N E, \text{Env}, K, t), rk), TT \rightsquigarrow \text{RApply}(t, node, rk), TT')} \\
\hline
\end{array}$$

## 7.2 Liveness

Evaluating will eventually produce a value

Decreasing on lexicalgraphic ordering on the replay stack do work

## 7.3 Performance

memory consumption is linear to amount of object with  $O(1)$  access cost

## 8 EVALUATION

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$$\begin{array}{c}
\frac{Query(tt, K) = (X, (\_, st)) \quad X! = K - 1}{(Apply(K, V, t), rk), TT \rightsquigarrow (st, RKApply V K rk), TT} \\
\\
\frac{Query(tt, K) = (K - 1, (Just KLeftK', \_)) \quad st = Apply(K', t + 1, t + 2) \quad node = (Just VLeftV, st)}{(Apply(K, V, t), rk), TT \rightsquigarrow RApply(t, node, rk), TT'} \\
\\
\frac{Query(tt, K) = (K - 1, (Just KRightK', \_)) \quad st = Apply(K', t + 1, t + 2) \quad node = (Just VRightV, st)}{(Apply(K, V, t), rk), TT \rightsquigarrow RApply(t, node, rk), TT'} \\
\\
\frac{Query(tt, K) = (K - 1, (Just KCaseEnvLNLNRK', \_)) \quad Query(tt, V) = (X, (\_, st)) \quad X! = V - 1}{(Apply(K, V, t), rk), TT \rightsquigarrow (st, RKCase V EnvLNLNRK' rk), TT} \\
\\
\frac{Query(tt, K) = (K - 1, (Just KCaseEnvLNLNRK', \_)) \quad Query(tt, V) = (V - 1, (Just VLeftX, \_)) \quad st = Apply(K', X, t + 1, t + 2)}{(Apply(K, V, t), rk), TT \rightsquigarrow RApply(t, node, rk), TT} \\
\\
\frac{Query(tt, K) = (K - 1, (Just KCaseEnvLNLNRK', \_)) \quad Query(tt, V) = (V - 1, (Just VRightX, \_)) \quad st = Apply(K', X, t + 1, t + 2)}{(Apply(K, V, t), rk), TT \rightsquigarrow RApply(t, node, rk), TT} \\
\\
\frac{Query(tt, K) = (K - 1, (Just KProd0EnvRK', \_)) \quad st = Eval(R, Env, t + 1, t + 2) \quad node = (Just KProd1V, st)}{(Apply(K, V, t), rk), TT \rightsquigarrow RApply(t, node, rk), TT'} \\
\\
\frac{Query(tt, K) = (K - 1, (Just KProd1LK', \_)) \quad st = Apply(K', t + 1, t + 2) \quad node = (Just VProdLV, st)}{(Apply(K, V, t), rk), TT \rightsquigarrow RApply(t, node, rk), TT'} \\
\\
\frac{Query(tt, K) = (K - 1, (Just KZroK', \_)) \quad Query(tt, V) = (X, (\_, st)) \quad X! = V - 1}{(Apply(K, V, t), rk), TT \rightsquigarrow (st, RKZro V K' rk), TT} \\
\\
\frac{Query(tt, K) = (K - 1, (Just KZroK', \_)) \quad Query(tt, V) = (V - 1, (VProdXY, \_)) \quad st = Apply(K', X, t + 1, t + 2)}{(Apply(K, V, t), rk), TT \rightsquigarrow RApply(t, node, rk), TT} \\
\\
\frac{Query(tt, K) = (K - 1, (Just KFstK', \_)) \quad Query(tt, V) = (X, (\_, st)) \quad X! = V - 1}{(Apply(K, V, t), rk), TT \rightsquigarrow (st, RKFst V K' rk), TT} \\
\\
\frac{Query(tt, K) = (K - 1, (Just KFstK', \_)) \quad Query(tt, V) = (V - 1, (VProdXY, \_)) \quad st = Apply(K', X, t + 1, t + 2)}{(Apply(K, V, t), rk), TT \rightsquigarrow RApply(t, node, rk), TT} \\
\\
\frac{Query(tt, K) = (K - 1, (Just KLetAEnvCK', \_)) \quad st = Eval(C, Env[A := V], K', t + 1) \quad node = (Nothing, st)}{(Apply(K, V, t), rk), TT \rightsquigarrow RApply(t, node, rk), TT'} \\
\\
\frac{Query(tt, K) = (K - 1, (Just KApp0EnvXK', \_)) \quad Query(tt, V) = (X', (\_, st)) \quad X'! = V - 1}{(Apply(K, V, t), rk), TT \rightsquigarrow (st, RKApply V Env XK' rk), TT} \\
\\
\frac{Query(tt, K) = (K - 1, (Just KApp0EnvXK', \_)) \quad Query(tt, V) = (V - 1, (ClosEnv'NE), \_), st = Eval(X, Env, t + 1, t + 2)}{(Apply(K, V, t), rk), TT \rightsquigarrow RApply(t, node, rk), TT} \\
\\
\frac{Query(tt, K) = (K - 1, (Just KApp1EnvNEK', \_)) \quad st = Eval(E, Env[N := V], K', t + 1) \quad node = (Nothing, st)}{(Apply(K, V, t), rk), TT \rightsquigarrow RApply(t, node, rk), TT'}
\end{array}$$