

$$E = \left\{ (x^i, w, y^j) \mid x^i, y^j \in LVar \right. \\ \left. w = \max \left\{ \begin{array}{l} \exists (\tau) : \text{Dep}(x^i, y^j, \tau, c) \end{array} \right\} \right\}$$

example: input x:

```

y=0;
i=0;
w=0;
z<q1;
if x=0 then
  (y < q2(z); i=x-1)
else
  (skip);
while i<neq x
  if i<neq x then w=0 else skip;
  i=i+1
  {w<q3(y,w); if i neq x-2 then w=0 else skip; i=i+1}
  
```

a trace τ_1, τ_2, τ_3
 $\tau_1: \tau_2, \tau_3$
 $\text{change } \tau_2 \Rightarrow \tau'_2 = \tau_2, \tau_3$

then another trace: $\tau_1 = w, \tau_2, \tau_3 = w, \tau_2$
 $\text{change } \tau_2 \Rightarrow \tau'_2 = w, v, \tau_3 = w, \tau_2$

+ trace s.t. $\# \text{trace} \geq \# \text{event} \geq \# \text{event where value of } w \text{ is changed}$.
 $\text{Dep}(\tau, \tau, w)$, # of occurrence time of $w \geq \# \text{of event where value of } w \text{ is changed}$.

then count: # trace τ s.t. $\text{Dep}(\tau, \tau, w)$.
 $| \text{Diff}(\tau_1, \tau_2, w) |$

\Rightarrow weight on edge (x^i, y^j, w) .
 $w = \max \{ | \text{Diff}(\tau_1, \tau_2, y^j) | + \tau_1, \tau_2, \text{s.t. } \text{Dep}(x^i, \tau_1, \tau_2, y^j) \}$

\Rightarrow 1. Dep on ω traces. redefine.

\Rightarrow ω Global Dep. ??

\Rightarrow add iteration id.

↳ possible unsound ??

then,

if cases H.t. observe the same
value, but in different iterations?
then intuitively no-dependency it is good.

\Rightarrow 2. $\text{Diff}(\tau_1, \tau_2, x^i)$

\Rightarrow $\#$
 $\text{Seq}(\tau_2, x^i) \triangleq \begin{cases} \tau_2 = \tau'_2 :: (x, i, y, z) \rightarrow \text{Seq}(\tau_2, \tau'_2, x^i) \\ \tau_2 = \tau'_2 :: (x, i, y, \alpha) \rightarrow \text{Seq}(\tau_2, \tau'_2, x^i) \\ \tau_2 = \{\} \rightarrow \{\} \\ \text{o.w.} \rightarrow \text{Seq}(\tau_2, x^i) \end{cases}$

without loop iteration id.

for $\text{Seq}(\tau_2, x^i)$, where $\text{Dep}(y^j, x^i, \tau_1, \tau_2)$.

case of $\text{Seq}(\tau_2) \neq \text{Seq}(\tau_2)$:

①. $\text{len}_{\tau_1} \neq \text{len}_{\tau_2} \Rightarrow \text{Diff}(\text{Seq}(\tau_1), \text{Seq}(\tau_2), x^i) = \{ k \mid \text{Seq}(\tau_1)[k] \neq \text{Seq}(\tau_2)[k] \cup \{ \text{len}_{\tau_1}, \text{len}_{\tau_1}+1, \dots, \text{len}_{\tau_2} \}$.

②. $\text{len}_{\tau_1} = \text{len}_{\tau_2} \Rightarrow \text{Diff}(\text{Seq}(\tau_1), \text{Seq}(\tau_2), x^i) = \{ k \mid \text{Seq}(\tau_1)[k] \neq \text{Seq}(\tau_2)[k] \}$

\Rightarrow potential issue: the different observed value is in different iterations.

case: if guard is affected by moving to different iteration, but it always executed the same time with same value.

then it doesn't matter which iteration it is in, then isn't dependence / adaptivity.

\Rightarrow for case ② where they have some len, it doesn't loss adaptivity if I observe the same value in the same location of $\text{Seq}(\tau_1)$ and $\text{Seq}(\tau_2)$

but where they actually comes from different loop iteration.

i.e. it is sound that only consider the diff of same location.

\Rightarrow for case ①. $\text{len}_{\tau_1} \neq \text{len}_{\tau_2}$.

in the ω seqs from $[0, \min(l_1, l_2)]$, it is still sound by just observe the different value in same location,

for the same reason as case ②.

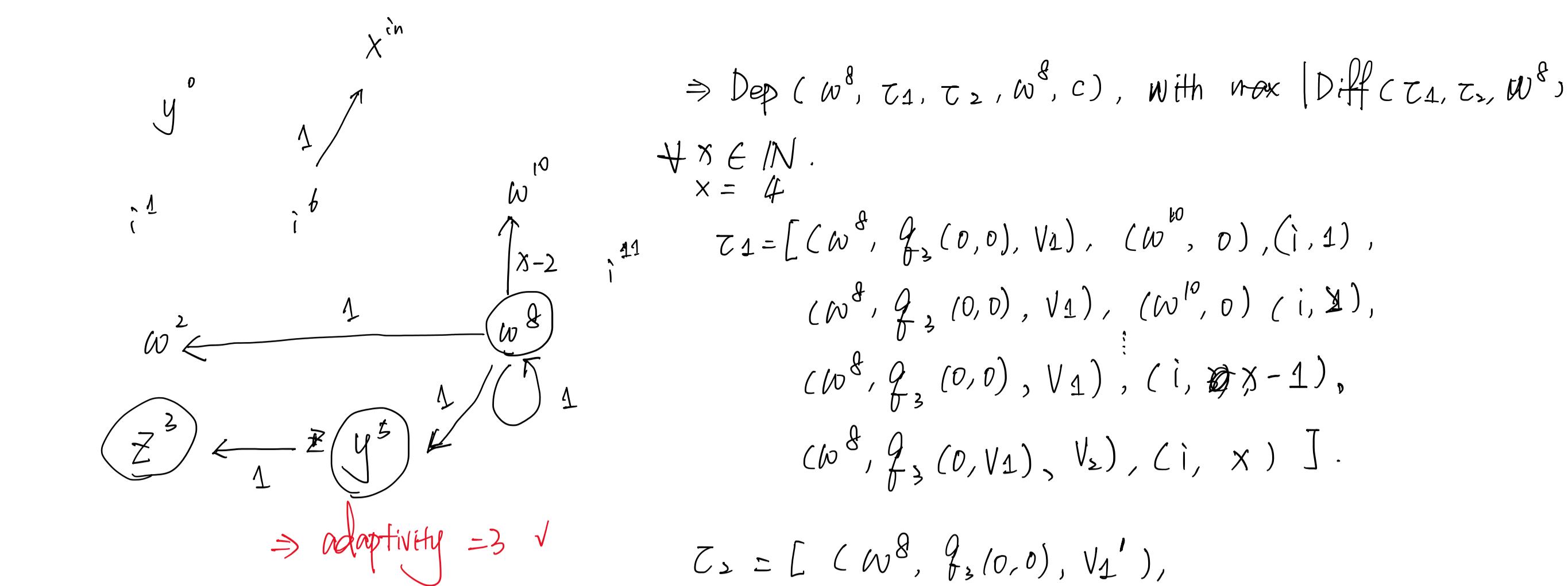
in the longer seq from $[\min(l_1, l_2), \max(l_1, l_2)]$.

case: $i=0$
 while $i < x$ do:
 $\quad i=i+1$
 $\quad y=g(x)$
 \Rightarrow when $x=100$
 \quad still safe.
 $\quad (y, 1, v_1) (y, 1, v_2) (y, 1, v_3) (y, 1, v_4) \dots$
 $\quad (y, 1, v_1') (y, 1, v_2') (y, 1, v_3') (y, 1, v_4) \dots$

$\Rightarrow Gy^*(i_0, x)$.

if in order to get different length, only way is by changing value affect guards.
 $(x, n, 1) (y, 1, v_1) \dots$
 \Rightarrow all the value are the same. then

$(x, 2n, 100), (y, 1, v_1) \dots$ adaptivity is actually 1.



\Rightarrow Dep Graph:

$\forall x^i \mid x^i \in LVar(C)$

$$\text{E}(\{ (x^i, y^j, w) \mid w = \max \{ | \text{Diff}(\tau_1, \tau_2, y^j) | \mid \# \tau_1, \tau_2, \text{s.t. } \text{Dep}(x^i, \tau_1, \tau_2, y^j) \} \} \}$$

$$\wedge w \geq 1 \wedge x^i, y^j \in LVar(C)$$

$\text{Diff}(\tau_1, \tau_2, x^i) \triangleq \{ k \mid (k=0, \dots, l_{\max} \wedge \text{Seq}(\tau_1, x^i)[k] \neq \text{Seq}(\tau_2, x^i)[k]) \}$

$\wedge (\text{Seq}(\tau_{\max}, x^i)[k+1] \neq \text{Seq}(\tau_{\max}, x^i)[k] \wedge k = l_{\min}, \dots, l_{\max})$

$\wedge l_{\min} = \min(|\text{Seq}(\tau_1, x^i)|, |\text{Seq}(\tau_2, x^i)|)$

$l_{\max} = \max(\dots)$

$\tau_{\max} = \text{longer Sequence. } \tau$

$\Rightarrow \text{Seq}_f(\tau, x^i) \triangleq \begin{cases} \text{seq}(\tau, x^i) \wedge \tau = \tau :: (x, i, y, \alpha) \\ \text{seq}(\tau, x^i) \wedge \tau = \tau' :: (x, i, y, \alpha) \\ \{\} \wedge \tau = \{\} \\ \text{seq}(\tau) \wedge \text{o.w.} \end{cases}$

$\text{Dep}(x^i, \tau_1, \tau_2, y^j, C) \triangleq$

$\exists \tau_1', \tau_2, \tau'$

$\tau_1 = (x, i, y, -), \tau_2 = (y, j, -, -)$

$\tau_1 \neq \tau_2$

$\langle C, \tau \rangle \rightarrow \langle C_1, \tau_1 \rangle :: \tau_1 \rightarrow \langle \text{skip}, \tau_1 :: \tau_1 + \tau_2 \rangle$

$\wedge \langle C_1, \tau_1 :: \tau_1' \rangle :: \tau_1' \rightarrow \langle \text{skip}, \tau_1' :: \tau_1 + \tau_2 \rangle$

$\text{Diff}(\tau_1, \tau_2, y^j) \neq \emptyset$.

\Rightarrow example:

↳ if they are different, then either y itself is used (through a chain) or y is used in evaluating y .

case: $i=0$
 while $i < x$:
 $\quad i=i+1$
 $\quad y=g(x)$
 \Rightarrow when $x=100$
 \quad still safe.
 $\quad (y, 1, v_1) (y, 1, v_2) (y, 1, v_3) (y, 1, v_4) \dots$

case: $i=0$
 while $i < x$:
 $\quad i=i+1$
 $\quad y=g(x)$
 \Rightarrow when $x=100$
 \quad still safe.
 $\quad (y, 1, v_1) (y, 1, v_2) (y, 1, v_3) (y, 1, v_4) \dots$

two approaches:
 ↳ prove there must exist value dependency from x to y or from other variable to y .

↳ define $|\text{Diff}|$ as $\{ \text{seq}(\tau_{\max})[m:n] \}$, w/o sound. set of unique values.

possible case: $(1, 0, 1, 0, 1, 0)$ I actually have no intuition does this adaptive or not.

\Rightarrow if it is, then $|\text{Diff}|$ defined as $\{ k \mid \text{seq}[k:D] \neq \text{Seq}[k:D]$ can capture.

but "intuitive adaptivity" is unclear.

Hence:
 $\max \{ | \text{Diff}(\tau_1, \tau_2, w^k) | \mid \# \tau_1, \tau_2, \text{Dep}(w^k, \tau_1, \tau_2, w^k, C) \} = 1$.

$\Rightarrow | \text{Diff}(\tau_1, \tau_2, w^k) | = 1$.

\Rightarrow Reduce the "flowsto" Over-approximate

I: eliminate Useless Data-Control Flow:

① Redundant If Branch.

case 1: Normalized Boolean Expr in Guard is constant.

\Rightarrow Remove another Block.

\Rightarrow Remove the data-control "flowsto"

of variables assigned inside the Block on Variables in guard.

case 2: 2 Blocks in two Branches are syntactically identity.

\Rightarrow Remove another Block.

\Rightarrow Remove the data-control "flowsto" from all Variables in guard.

to variables assigned inside the Block.

$x \leftarrow 0.$

\Rightarrow if $(x < 3, [y=1], [y=2])$

$\exists v, \forall c, <c, b> \Downarrow v$

if $c, b, [y=1], [y=1]$.

2 branches are the same

② Redundant While loop.

Case 1: Normalized Boolean Expr in Guard is constant false.

\Rightarrow Remove the Block of the while Body

\Rightarrow Remove the data-control "flowsto" from all Variables in guard.

to variables assigned inside the Block.

$x \leftarrow 0$

while $x > 0,$

$y = 1.$

Same as if-case 1. $\exists v, \forall c, <c, b> \Downarrow v.$

③ Redundant variables in Guard.

Forb of guard in every if and while command.

$\forall x \in \text{VAR}(b), x \notin \text{VAR}(\text{NF}(b))$

remove all the data-control "flowsto" from x to variables in the Body Block

$x \leftarrow g(0)$

if $c, [x-x=0], [y=1], [y=2]$.

Variable disappeared from $\text{VAR}(b)$, or $\text{VAR}(e)$ or $\text{VAR}(a)$ by normalization



$x \leftarrow g(0)$

$y \leftarrow g(x - x).$

same as boolean expression.

remove dependency after normalization.

\downarrow

II. Remove redundant data-value "flowsto".

for the e and ψ in every assignment command. $y \leftarrow e$ and $y \leftarrow \psi$

$\forall x \in \text{VAR}(e), x \in \text{VAR}(\text{NF}(e)), \text{ or } x \in \text{VAR}(\psi) \wedge x \notin \text{VAR}(\text{NF}(\psi))$

Remove the data-value "flowsto" from x to y

$x \leftarrow g(0)$

$y \leftarrow g(x - x).$

same as boolean expression.

remove dependency after normalization.

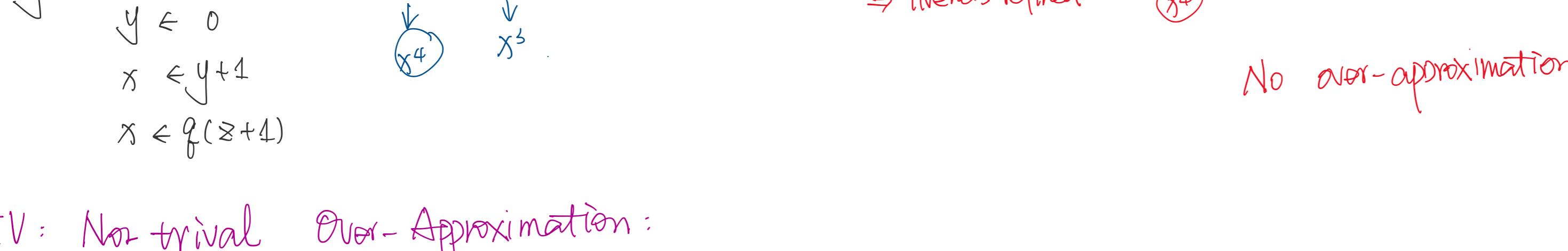
\downarrow

III: Non-trivial liveness Analysis. \Rightarrow variable Dead.

\Rightarrow Cannot further Optimize.

example 2: Event Rewriting occurs in a longer adaptivity chain,

and removing the other vertex is safe, keeping both flow doesn't lead to over-approximation.



IV: Non-trivial Over-Approximation:

path Sensitive syntactically but not semantically:

\Rightarrow 2 if command have the same guard, they take the same branch always.

the adaptivity Expected by $\max \left\{ \begin{array}{l} \text{Adap}_{\text{truebranch1}} + \text{Adap}_{\text{truebranch2}} \\ \text{Adap}_{\text{f-branch1}} + \text{Adap}_{\text{f-branch2}} \end{array} \right\}$

the Adapt program = $\max \left\{ \begin{array}{l} \text{Adap}_{\text{truebranch1}} + \text{Adap}_{\text{truebranch2}} \\ \text{Adap}_{\text{f-branch1}} + \text{Adap}_{\text{f-branch2}} \\ \text{Af-branch1} + \text{Af-branch2} \\ \text{Af-branch1} + \text{Af-branch2} \end{array} \right\}$

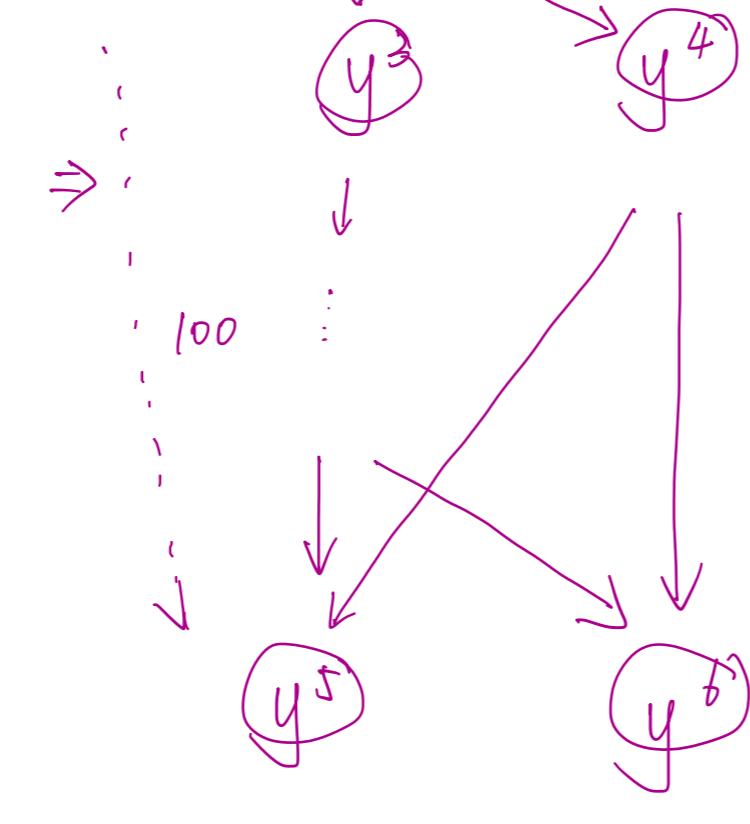
eg: input a: $y \leftarrow g(0)$

if $a > 0$:
 $y \leftarrow g^{100}(y)$

else:
 $y \leftarrow g(y).$

if $a > 0$:
 $y \leftarrow g(y)$

else:
 $y \leftarrow g^{100}(y).$



longest path: 201. if $100 = n \Rightarrow 2n+1$

Expected Adapt: 102. $\Rightarrow n+2.$

Limitation-I-Flowsto-Overapprox

Friday, May 20, 2022 10:28 AM

in the CFG Analysis: in $x = y - y$.

Semantic dependency $\text{Dep}(x^i, y^i, c)$ is traditionally

- ① over-approximated syntactically by $\text{VAR}(a_e)$ and $\text{VAR}(b)$.

We can improve by normalize the expression syntactically.

$$\text{VAR}(a_1 - a_2) = \{y \mid y \in \text{VAR}(a_1) \cup \text{VAR}(a_2) \wedge \text{NF}(a_1) \neq \text{NF}(a_2)\}$$

$$\text{VAR}(b_1 = a_2) = \{y \mid y \in \text{VAR}(a_1) \cup \text{VAR}(a_2) \wedge \text{NF}(a_1) \neq \text{NF}(a_2)\}$$

$x = 1$

if $x = x$

then $y = 1$

else $y = 2$.

if $x = 1$:

then $y = 1$

else $y = 2$

- ② also over-approximated syntactically by conditional commands:

$x = 2$

if $x > 0$

then $y = 1$

else $y = 2$.

Example-II

Tuesday, July 12, 2022

4:12 PM

Example II :

$\bar{z} = f_{\text{univ}}(0)$] ⁰
 $j = k^2 + 1$; ¹
 while $j > 0$ ²
 { $j \leftarrow j - 1$ ³
 $z \leftarrow f_{\text{univ}}(\bar{z} * x[j])$ ⁴ } ⁵
 if $k \leq 2$ ⁵
 then { $y \leftarrow f_{\text{univ}}(\bar{z})$ } ⁶ ⁶
 else { $y \leftarrow 0$; } ⁷ ⁷
 $i \leftarrow k$ ⁸
 while $i > 0$ ⁹ do
 { $i \leftarrow i - 1$; } ¹⁰
 $y \leftarrow f_{\text{univ}}(y)$; ¹¹ ¹¹

in previous definition.
⇒

The diagram illustrates a directed graph with nodes labeled by variables:

- w : 1
- z^0
- y^t
- y^{11}
- y^7
- ω : $\tau \rightarrow \begin{cases} 1 & (\tau k) \leq 2 \\ 0 & (\tau k) > 2 \end{cases}$

Directed edges are as follows:

- w : 1 has a curved arrow pointing to z^0 .
- z^0 has a curved arrow pointing to w : 1.
- w : 1 has a curved arrow pointing to y^t .
- y^t has a curved arrow pointing to w : $\tau \rightarrow \begin{cases} 1 & (\tau k) \leq 2 \\ 0 & (\tau k) > 2 \end{cases}$.
- w : $\tau \rightarrow \begin{cases} 1 & (\tau k) \leq 2 \\ 0 & (\tau k) > 2 \end{cases}$ has a curved arrow pointing to y^{11} .
- y^{11} has a curved arrow pointing to w : $\tau \rightarrow \begin{cases} 1 & (\tau k) \leq 2 \\ 0 & (\tau k) > 2 \end{cases}$.
- y^{11} has a curved arrow pointing to y^7 .
- y^7 has a curved arrow pointing to y^t .

I v consider both vertices weight & edge weight. v

restricted walk : To →

$$\left\{ \begin{array}{l} \text{edge seq : } (e_1, \dots, e_n) \quad \text{s.t. } \text{cnt}(e_i) \leq w_{ei}(T_0) \\ \text{with Vertices seq } (v_1, \dots, v_{n+1}). \quad \text{cnt}(v_i) \leq w_{vi}(T_0) \end{array} \right.$$

$$\text{flen}(k, c) : \mathbb{Z} \rightarrow \mathbb{N}$$

$$\Rightarrow \text{fun}(k, c)(\tau_0) = \#\{1 \mid v_i \in Q \vee (c) \wedge v_i \in k(\omega) \wedge k \in \text{WALK}(c)\}.$$

$$\Rightarrow A(c) : \tau_0 \rightarrow N$$

$$A(c) = \tau_0 \rightarrow \max \{ q_{\text{len}}(k, c)(\tau_0) \mid k \in \mathcal{W}(c) \}.$$

 better

Example Limitation:

still the example of limitation:

Simplified : Multi Rounds Single (k) :

[j < k]

$[z \leftarrow \text{query}^{\langle 0 \rangle}]^1$

while $[j > 0]^2$ do

$$\{ [y \leftarrow f(\text{key}(y * z))];$$

if ($[j \neq 2] ^ 4$
 $[j < j - 1] ^ 7$)

‘*It’s a good place to go to when you’re feeling down.*’

$\omega: \mathbb{N} \rightarrow \mathbb{Z}$
 $\omega: \mathbb{Z} \rightarrow \mathbb{Z}$

$\omega: \begin{cases} (\mathbb{Z}^k) - 1 & \mathbb{Z}^k \geq 2 \\ (\mathbb{Z}^k) & \mathbb{Z}^k < 2 \end{cases}$

\Rightarrow adaptivity. indeed $\bar{z} \rightarrow \begin{cases} > & (\bar{z}_k) = 1 \\ \geq & (\bar{z}_k) \geq 2 \end{cases}$

$$A(c) = \tau \Rightarrow (\tau k) + 1$$

under New Definition.

The diagram illustrates the composition of three functions:

- $w: T \rightarrow 1$ (bottom right)
- $w: T \rightarrow \{1, 2\}$ (top right, enclosed in a red box)
- $w: T \rightarrow T^k$ (middle right, enclosed in a red box)
- z^4 (top left)
- y^{3k} (bottom left)

Arrows indicate the flow of data from the bottom function up to the middle one, and from the middle one up to the top one. The top function is enclosed in a red box, and its output is labeled $(T^k) = 1$ and $(T^k) \geq 2$.

$$\Rightarrow \text{Acc}(\tau) = \tau \rightarrow \left\{ \begin{array}{ll} 1 & (\tau k) = 0 \\ \geq & (\tau k) = 1 \\ \rightarrow & (\tau k) \geq 2 \end{array} \right.$$

New Over-Approximate Examples

Monday, July 11, 2022

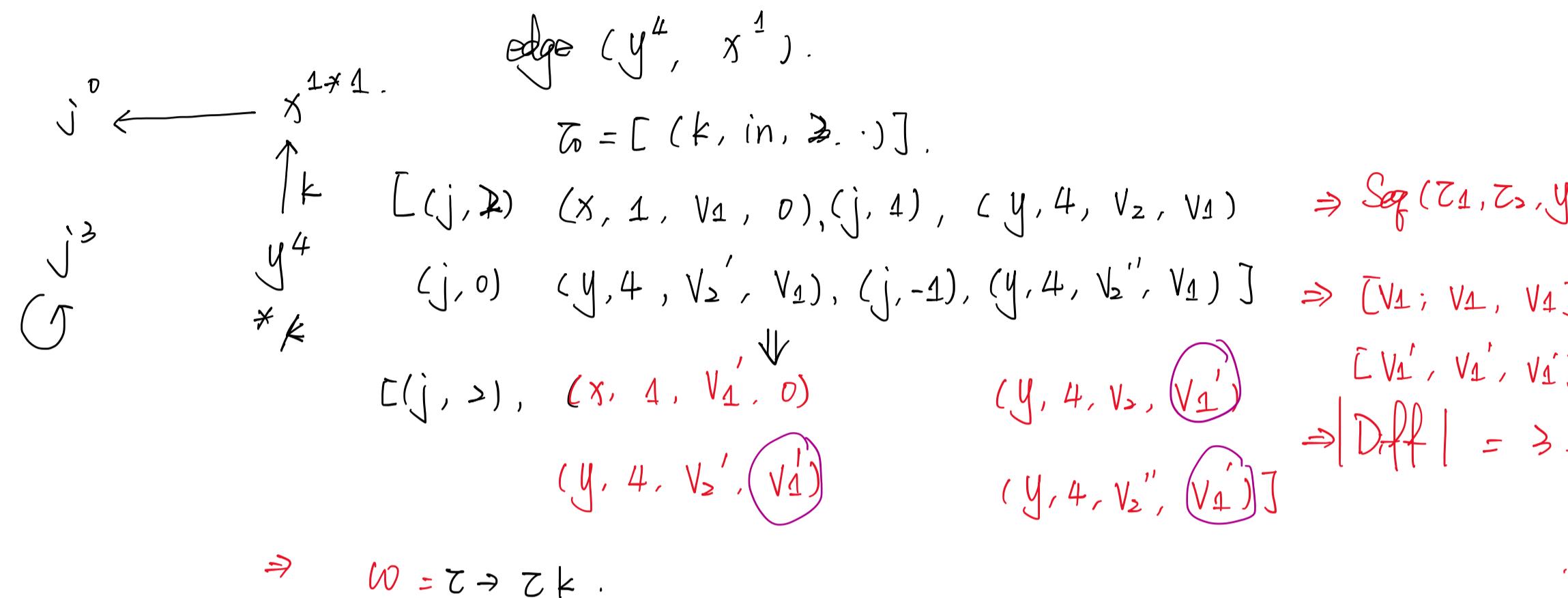
11:21 AM

- ⇒ over-approximate dependency between variable outside loop. v.s. inside.
- ⇒ change a constant assignment for variable outside loop.
will cause sequence of difference for variable inside loop. if $y \leftarrow x$.
- ⇒ dependency actually 1, but edge weight is iteration #.

Example:

```

 $[j < k]$ 
 $[x \leftarrow \text{query}(0)]^1$ 
while  $[j \geq 0]^2$  do
   $[j \leftarrow j - 1]^3$ 
   $[y \leftarrow \text{query}(x)]^4$ .
  
```



⇒ 2 ways:

I ✓ consider both vertices weight & edge weight. v.

restricted walk : $\tau_0 \rightarrow$

$$\left\{ \begin{array}{l} \text{edge seq} : (e_1, \dots, e_n) \\ \text{s.t. } \text{cnt}(e_i) \leq w_{ei}(\tau_0) \end{array} \right.$$

$$\text{with Vertices seq} (v_1, \dots, v_{n+1}). \quad \text{cnt}(v_i) \leq w_{vi}(\tau_0).$$

$$f_{len}(k, c) : \tau_0 \rightarrow \mathbb{N}$$

$$\Rightarrow f_{len}(k, c)(\tau_0) = \#\{i \mid v_i \in Q \vee (c) \wedge v_i \in k(\tau_0) \wedge k \in \text{WALK}(c)\}$$

$$\Rightarrow A(c) : \tau_0 \rightarrow \mathbb{N}$$

$$A(c) = \tau_0 \rightarrow \max \{ f_{len}(k, c)(\tau_0) \mid k \in \text{WALK}(c) \}$$

\downarrow
better?

both works.

⇒ better.

$$\text{II: } \Rightarrow |\text{Diff}| = 1.$$

$$\text{Diff}_{\text{seq}}(\tau_1, \tau_2, x^l) \triangleq$$

$$\left\{ \begin{array}{l} \text{Seq}_1[k] \neq \text{Seq}_2[k] \text{ or } \text{Seq}_1[0] \neq \text{Seq}_2[0] \text{ or } \text{Seq}_1 = [] \\ (k=1 \dots \text{len}(\text{Seq}_2)), \quad \text{Seq}_2[k-1] \neq \text{Seq}_1[k] \end{array} \right.$$

$$\cup \{ \text{Seq}_1[0] \mid \text{Seq}_2 = [] \}$$

$$\text{Diff} \triangleq (\text{len}(\text{Seq}_1) \neq \text{len}(\text{Seq}_2)) \vee (\text{Seq}_1[0] \neq \text{Seq}_2[0])$$

$$\Rightarrow \text{Diff}_{\text{seq}}(\tau_1, \tau_2, x^l) \triangleq$$

$$\left\{ \begin{array}{l} v \mid \text{Seq}_1[0] \neq \text{Seq}_2[0] \quad (\text{Seq}_1 \neq [] \wedge \text{Seq}_2 = []) \\ v \mid \text{Seq}_2[k] \neq \text{Seq}_1[k] \quad (\text{Seq}_2 \neq [] \wedge \text{Seq}_1 = [] \wedge \text{Seq}_2[k] \neq \text{Seq}_1[k+1], k=0, \dots, \text{len}(\text{Seq}_2)) \\ v \mid \text{Seq}_2[k] \quad (\text{Seq}_2 \neq [] \wedge \text{Seq}_1 \neq [] \wedge \text{Seq}_1[0] \neq \text{Seq}_2[0], \text{Seq}_1[k] \neq \text{Seq}_2[k+1], k=0, \dots, \text{len}(\text{Seq}_1)) \end{array} \right.$$

$$\Rightarrow |\text{Diff}(\tau_1, \tau_2, y)| \text{ in example is 1.} \Rightarrow \text{weight of edge } \tau \rightarrow 1 \not\rightarrow$$

but it causes over approximate still.

because even $\text{Seq}[k+1] \neq \text{Seq}[k]$, the variation still doesn't come from modification of variable outside of loop.

Dep Graph:

 $\forall i \mid x^i \in LV(C)$

$$\exists \{ (x^i, y^j), w \mid w = \max \{ |Diff(\tau_1, \tau_2, y^j)| \mid \forall \tau_1, \tau_2, \text{st. } Dep(x^i, \tau_1, \tau_2, y^j) \} \}$$

functional

$$\Rightarrow \left\{ \begin{array}{l} w: \tau_0 \rightarrow \mathbb{N} \wedge x^i, y^j \in LV(C) \wedge \\ w(\tau_0) = \max \{ |Diff(\tau_1, \tau_2, y^j)| \mid \forall \tau_1, \tau_2, \text{st. } Dep(x^i, y^j, \tau_1, \tau_2, \tau_0, C) \} \end{array} \right\}$$

$$Diff(\tau_1, \tau_2, x^i) \triangleq \{ k \mid (k=0, \dots, l_{min} \wedge Seq(\tau_1) x^i[k] \neq Seq(\tau_2) x^i[k]) \}$$

$$\wedge (Seq(\tau_{max}, x^i)[k] \neq Seq(\tau_{max}, x^i)[k] \wedge k=l_{min}, \dots, l_{max})$$

$$\wedge l_{min} = \min(|Seq(\tau_1, x^i)|, |Seq(\tau_2, x^i)|)$$

$$l_{max} = \max(\dots)$$

$$\tau_{max} = \text{longer Sequence. } \tau$$

$$\Rightarrow Seq(\tau, x^i) \triangleq \begin{cases} \text{seq}(\tau, x^i) := \tau & \tau = \tau' : (x, i, y, \dots) \\ \text{seq}(\tau, x^i) := \tau & \tau = \tau' : (x, i, y, \dots) \\ \square & \sigma = \square \\ \text{seq}(\tau, x^i) & \text{o.w.} \end{cases}$$

$$Dep(x^i, \tau_1, \tau_2, y^j, C) \triangleq$$

$$\exists \epsilon_1, \tau_1, \tau_2 \mid \epsilon_1 = (x, i, \dots, \dots), \epsilon_2 = (y, j, \dots, \dots)$$

$$\epsilon_1 \neq \epsilon_2$$

$$\langle C, \tau_0 \rangle \rightarrow \langle C_1, \tau_1 \rangle \vdash \epsilon_1 \rightarrow \langle \text{skip}, \tau_1 \rangle \vdash \epsilon_1 + \tau_1$$

$$\wedge \langle C_1, \tau_1 \rangle \vdash \epsilon_2 \rightarrow \langle \text{skip}, \tau_1 \rangle \vdash \epsilon_2 + \tau_1$$

$$Diff(\tau_1, \tau_2, y^j) \neq \emptyset$$

$$Dep(x^i, y^j, \tau_1, \tau_2, \tau_0, C) \triangleq$$

$$\exists \tau_0, \epsilon_1, \epsilon_2 \mid \epsilon_1 = (x^i, i, \dots, \dots), \epsilon_2 = (y^j, j, \dots, \dots), \epsilon_1 \neq \epsilon_2$$

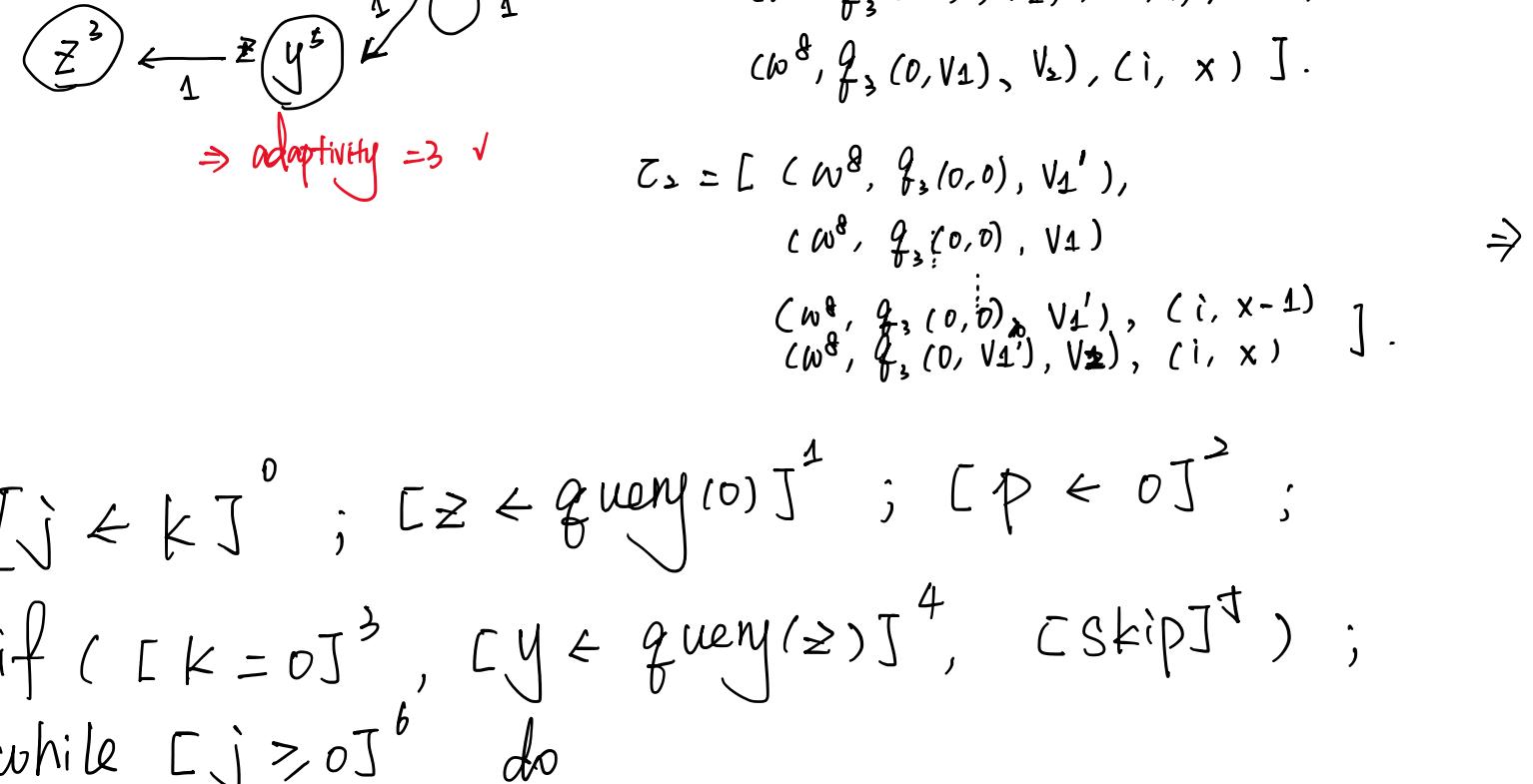
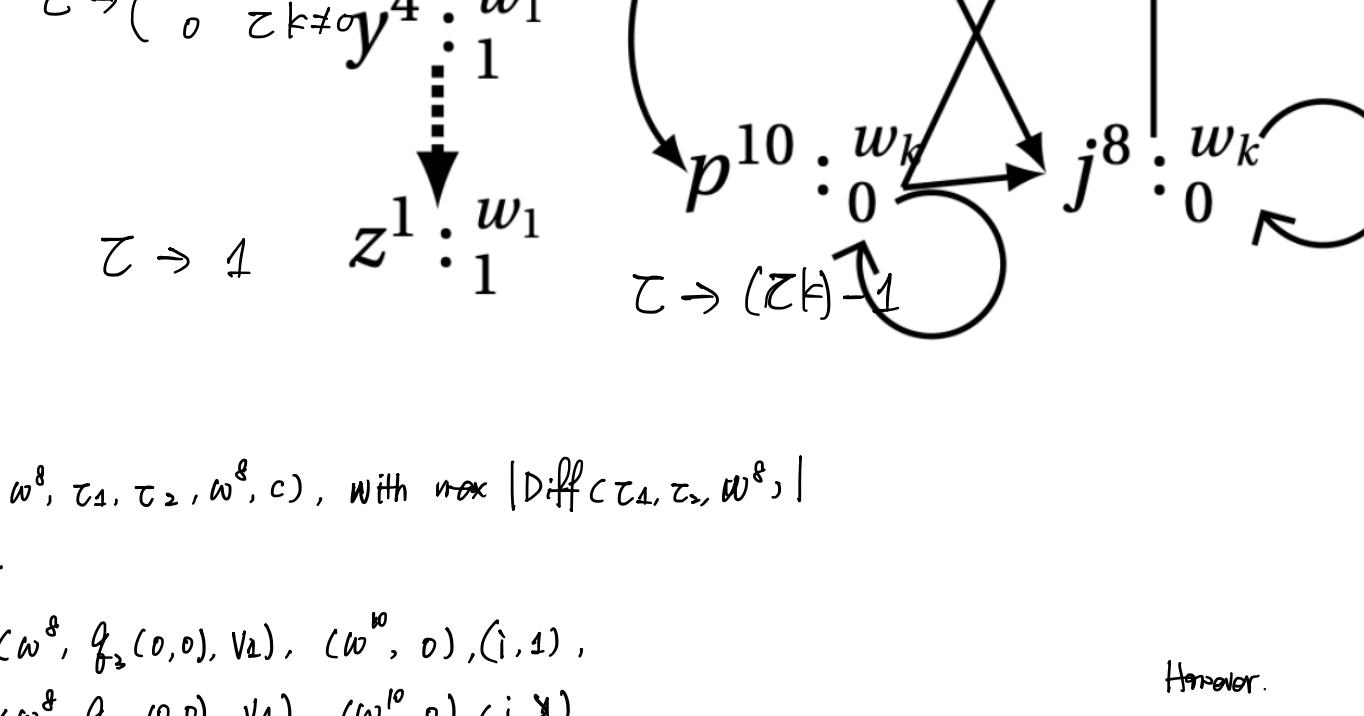
$$\langle C, \tau_0 \rangle \rightarrow \langle C_1, \tau_1 \rangle \vdash \epsilon_1 \rightarrow \langle \text{skip}, \tau_1 \rangle \vdash \epsilon_1 + \tau_1$$

$$\langle C_1, \tau_1 \rangle \vdash \epsilon_2 \rightarrow \langle \text{skip}, \tau_1 \rangle \vdash \epsilon_2 + \tau_1$$

$$\wedge Diff(\tau_1, \tau_2, y^j) \neq \emptyset.$$

Example :

```
multipleRoundsSingle(k)
[j - 0]^0; [z - query(0)]^1; [p - 0]^2;
if ([k = 0]^0, [y - query(z)]^4, [skip]^5);
while ([j ≠ k]^0 do
  ([p - query(x[y] + p)]^7; [j - j + 1]^8
  if ([j ≠ k - 2]^9, [p - 0]^10, [skip]^10));
```



$$\Rightarrow \begin{aligned} & [j \leq k^0; [z \leq f^{query}(0)]^4; [p \leq 0]^2; \\ & \text{if } ([k = 0]^3, [y \leq f^{query}(z)]^4, [\text{skip}]^5); \\ & \text{while } [j \geq 0]^6 \text{ do} \\ & \quad ([p \leq f^{query}(y + p)]^7; \\ & \quad [j \leftarrow j - 1]^8; \\ & \quad \text{if } ([j \neq k - 2]^9, [p \leq 0]^10, [\text{skip}]^10); \end{aligned}$$

New Dep graph :

$$\Rightarrow \tau_0 = [(k, in, \geq, \cdot)]$$

$$\text{edge: } (y^4, z^4)$$

$$\tau_1 = [(p, \neq, v_1, 0), (j, \geq, 2, \cdot), (j \neq 1, \text{true}, \cdot), (p, 10, 0, \cdot),$$

$$(j \geq 0, \text{true}), (p, \neq, v_1', 0), (j, \neq 1, \text{false}), \dots]$$

$$(j \geq 0, \text{true}), (p, \neq, v_1'', v_1'), (j, 0), (j \neq 1, \text{true}), (p, 10, 0).$$

$$(j \geq 0, \text{true}), (p, \neq, v_1''', v_1'), (j, -1), (j \neq 1, \text{true}), (p, 0, 0),$$

$$(j \geq 0, \text{false})].$$

$$\Rightarrow Seq(\tau_1, \tau_2, p^7) = \langle \frac{v_1'}{v_2 \cdot 0} \rangle \Rightarrow |Diff| = 1.$$

$$\text{if } \tau_0 = [(k, in, \geq, 0)] \quad \tau_2 = [(k, in, 0, \cdot)] \quad |Diff| = 0.$$

$$\Rightarrow \tau \rightarrow \begin{cases} 1 & \tau_0 k \geq 2 \\ 0 & \tau_0 k \leq 1. \end{cases}$$

$$\text{edge } (p^7, p^{10}) :$$

$$\tau_0 k = [(k, in, \geq, \cdot)].$$

$$\Rightarrow \tau_2 = [\dots (p, 10, 1, \cdot),$$

$$\dots (p, \neq, v_1', 1) (j, 4)$$

$$(p, \neq, v_1'', v_1') (j, 0) (p, 0)$$

$$(p, \neq, v_1''', v_1') (j, -4) (p, 0) \dots]$$

$$\Rightarrow \tau \rightarrow \begin{cases} 1 & \tau_0 k \geq 1 \\ 0 & \tau_0 k = 0. \end{cases}$$

$$\Rightarrow \text{adaptivity} : \begin{cases} 2 & \tau_0 k = 0 \\ 0 & \tau_0 k = 1 \\ 1 & \tau_0 k \geq 2 \end{cases}$$

\Rightarrow the over-approximation for dependency depth between variables inside loop.

caused from weight, & Dependency based on different traces.

Motivating Example

Friday, June 24, 2022 12:36 PM

$$E = \left\{ \begin{array}{l} x^i, w, y^j \mid x^i, y^j \in LV(c) \\ w = \max \{ g(\tau) \mid \text{Dep}(x^i, y^j, \tau, c) \} \end{array} \right\}$$

example: input x :

```

y=0;
i=0;
w=0;
z< q1;
if x=0
  {y< q2(z); i=x-1}
else
  (skip);
while i< neg x
  {w< q3(y); if i neq x-2 then w=0 else skip; i=i+1}
  
```

a trace τ_1, τ_2, τ_3 :
 $\tau_1 = w=2, \tau_2 = w=2, \tau_3 = w=2$
change $w \neq w$ $\Rightarrow \tau_1' = \tau_2, \tau_3'$

then another trace: $\epsilon_1 = w=2, \epsilon_2 = w=2, \epsilon_3 = w=2$
change $w \neq w$ $\Rightarrow \epsilon_1' = w=2, \epsilon_2 = w=2, \epsilon_3' = w=2, \tau_3'$

+ trace. s.t. $\text{Dep}(w, \tau, w)$, # of occurrence time of $w \geq$ # of start where value of w is changed.

then. count: + trace τ s.t. $\text{Dep}(w, \tau, w)$.
 $| \text{Diff}(\tau_1, \tau_2, w^*) |$

\Rightarrow weight on edge (x^i, y^j, w) :
 $w = \max \{ | \text{Diff}(\tau_1, \tau_2, y^j) | + \# \tau_1, \tau_2, \text{s.t. } \text{Dep}(x^i, \tau_1, \tau_2, y^j) \}$

\Rightarrow 1. Dep on ω traces. define.

\Rightarrow 2. Control Dep. ??

\Rightarrow add iteration id.
in order to know the value changed in the ω traces are in the same iteration.

then, if cases τ deserve the same value, but in different iterations?
then intuitively no-dependency. it's good.

\Rightarrow 3. $\text{Diff}(\tau_1, \tau_2, x^i)$

\Rightarrow ~~case~~
 $\text{seq}(\tau_1) x^i \triangleq \begin{cases} \tau_1 = \tau_1' :: (x, i, v, \alpha) \rightarrow \text{seq}(\tau_1') :: v \\ \tau_1 = \tau_1' :: (x, i, v, \alpha) \rightarrow \text{seq}(\tau_1') :: v \\ \tau_1 = \tau_1' :: (x, i, v, \alpha) \rightarrow \text{seq}(\tau_1') :: v \\ \text{o.w.} \rightarrow \text{seq}(\tau_1') \end{cases}$

without loop iteration id.

for $\text{seq}(\tau_1) x^i$, where $\text{Dep}(y^j, x^i, \tau_1, \tau_2)$.

case of $\text{seq}(\tau_1) \neq \text{seq}(\tau_2)$:

①. $\text{len}_1 \neq \text{len}_2 \Rightarrow \text{Diff}(\text{seq}(\tau_1), \text{seq}(\tau_2), x^i) = \{ k \mid \text{seq}(\tau_1)[k] \neq \text{seq}(\tau_2)[k] \vee k \in \text{len}_1, \text{len}_2+1, \dots, \text{len}_2 \}$.

②. $\text{len}_1 = \text{len}_2 \Rightarrow \text{Diff}(\text{seq}(\tau_1), \text{seq}(\tau_2), x^i) = \{ k \mid \text{seq}(\tau_1)[k] \neq \text{seq}(\tau_2)[k] \}$

\Rightarrow potential issue: the different observed value is in different iterations.

case: if y is affected by moving to different iteration, but it always executed the same time with same value.

then it doesn't matter which iteration it is in, thus isn't dependence / adaptivity.

\Rightarrow for case ②, where they have same len, it doesn't loss adaptivity if I observe the same value in the same location of $\text{seq}(\tau_1)$ and $\text{seq}(\tau_2)$
but where they actually comes from different loop iteration.

i.e. it's sound that only count the diff of some iteration.

\Rightarrow for case ③: $\text{len}_1 \neq \text{len}_2$.

in the ω seqs from $[0, \min(l_1, l_2)]$, it is still sound by just observe the different value in some location,

for the same reason as case ①.

in the longer seq from $[\min(l_1, l_2), \max(l_1, l_2)]$.

case: $i=0$
while $i < x$:
 $i = i+1$
 $y = f(a)$

\Rightarrow when $x=100$
 \quad still safe
 $\quad (y, 1, v_1) (y, 2, v_2) (y, 3, v_3) (y, 4, v_4) \dots$
 $\quad (y, 1, v_1) (y, 2, v_2) (y, 3, v_3) (y, 4, v_4) \dots$

$\Rightarrow Gy^*(inf, s)$.

if in order to get different length, only way is by changing value affect y .
 τ_1, τ_2, τ_3 : $(y, 1, v_1) (y, 2, v_2) \dots$
 \Rightarrow all the value are the same. then adaptivity is actually 1.

guard variable
 $\frac{\text{is used in evaluating } y}{\text{or } y \text{ itself is used through a chain}} \rightsquigarrow$
 \Rightarrow if they are different, then either

$i=0$
while $i < x$
 $w = f(w)$
 if $i > 3$, $w=0$, skip
 $i=i+1$

Dip: infinite.

Case: $i=0$
while $i < x$:
 $w = f(w)$
 if $i > 3$, $w=0$, skip. $x=100: (w, 1, f(1)), (w, 2, f(2)), (w, 3, f(3)), (w, 4, f(4)) \dots$

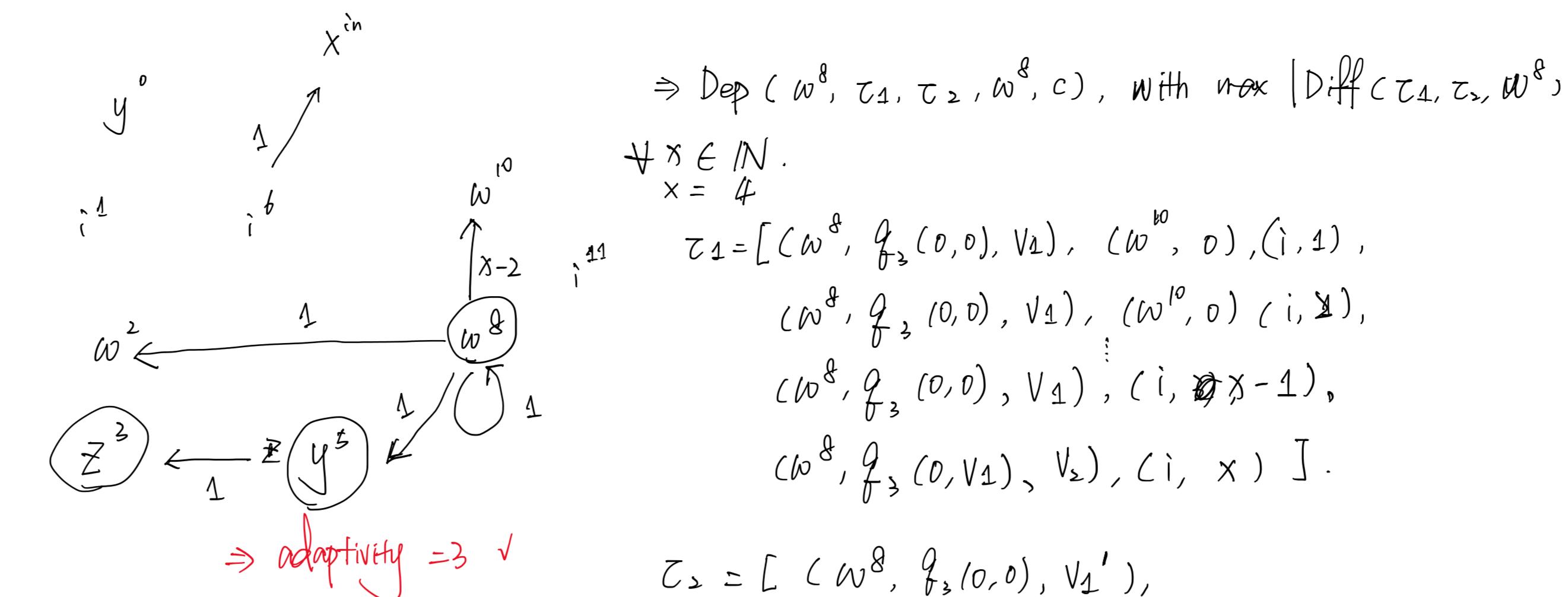
$i=i+1$

two approaches:
① prove there must exist value dependency from x to y or from other variable to y .

② define $|\text{Diff}|$ as $\{ \text{seq}(\tau_{\text{max}})[\min : \max] \}$. \rightsquigarrow sound. set of unique values.
possible case: $(1, 0, 1, 0, 1, 0)$ I actually have no intuition does this adaptive or not.

\Rightarrow if it is, then $|\text{Diff}|$ defined as $\{ k \mid \text{seq}[k-1] \neq \text{seq}[k] \}$ can capture.

but "intuitive adaptivity" is unclear.



\Rightarrow $\text{Dep}(w^8, \tau_1, \tau_2, w^8, c)$, with $\max |\text{Diff}(\tau_1, \tau_2, w^8)|$
 $\forall x \in \mathbb{N}$.
 $\tau_1 = [(w^8, f_1(0,0), V_1), (w^10, 0), (i, 1), (w^8, f_3(0,0), V_1), (w^10, 0), (i, 2), (w^8, f_3(0,0), V_1), (i, 3-1), (w^8, f_3(0,0), V_1), (i, x)]$.

$$\max \{ |\text{Diff}(\tau_1, \tau_2, w^8)| \mid \text{Dep}(w^8, \tau_1, \tau_2, w^8, c) \} = 1.$$

\Rightarrow Diff Graph:

$\forall i \mid x^i \mid x^i \in LV(c)$

$\text{EGF} \{ (x^i, y^j, w) \mid w = \max \{ |\text{Diff}(\tau_1, \tau_2, y^j)| \mid \# \tau_1, \tau_2, \text{s.t. } \text{Dep}(x^i, \tau_1, \tau_2, y^j) \} \}$

$\text{Diff}(\tau_1, \tau_2, x^i) \triangleq \{ k \mid (k=0, \dots, l_{\min} \wedge \text{Seq}(\tau_1) x^i[k] \neq \text{Seq}(\tau_2) x^i[k]) \wedge$

$\wedge (\text{Seq}(\tau_{\max}, x^i)[k-1] \neq \text{Seq}(\tau_{\max}, x^i)[k] \wedge k=l_{\min}, \dots, l_{\max}) \wedge$

$\wedge l_{\min} = \min(|\text{Seq}(\tau_1, x^i)|, |\text{Seq}(\tau_2, x^i)|)$

$l_{\max} = \max(\dots)$

$\tau_{\max} = \text{longer Sequence. } \tau$

$\Rightarrow \text{Seq}(\tau, x^i) \triangleq \begin{cases} \text{seq}(\tau', x^i) :: v & \tau = \tau' :: (i, v, \alpha) \\ \text{seq}(\tau', x^i) :: v & \tau = \tau' :: (i, v, \alpha) \\ \square & \tau = \square \\ \text{seq}(\tau_1') & \text{o.w.} \end{cases}$

$\text{Dep}(x^i, \tau_1, \tau_2, y^j, c) \triangleq$

$\exists \tau_1', \tau_2', \tau_0'$

$\epsilon_1 = (x, i, v_-, -), \epsilon_2 = (y, j, -, -)$

$\epsilon_1 \neq \text{start } \epsilon_1'$

$\langle C, \tau_0 \rangle \rightarrow \langle C_1, \tau_1' :: \epsilon_1 \rangle \rightarrow \langle \text{skip}, \tau_0' :: \epsilon_1 + \tau_1 \rangle$

$\wedge \langle C_1, \tau_0' :: \epsilon_1' \rangle \rightarrow \langle \text{skip}, \tau_1' :: \epsilon_1 + \tau_1 \rangle$

$\text{Diff}(\tau_1, \tau_2, y^j) \neq \emptyset$.

\Rightarrow example:

\Rightarrow

if in order to get different length, only way is by changing value affect y .

τ_1, τ_2, τ_3 : $(y, 1, v_1) (y, 2, v_2) \dots$

\Rightarrow all the value are the same. then adaptivity is actually 1.

guard variable
 $\frac{\text{is used in evaluating } y}{\text{or } y \text{ itself is used through a chain}}$

\rightsquigarrow

\Rightarrow if they are different, then either

y itself is used through a chain

or y is used in evaluating y .

\Rightarrow if y is used in evaluating y , then adaptivity is 1.

\Rightarrow if y itself is used through a chain, then adaptivity is 1.

\Rightarrow if y is used in evaluating y and y itself is used through a chain, then adaptivity is 1.

\Rightarrow if y is used in evaluating y and y itself is not used through a chain, then adaptivity is 1.

\Rightarrow if y itself is used through a chain and y is not used in evaluating y , then adaptivity is 1.

\Rightarrow if y is not used in evaluating y and y itself is not used through a chain, then adaptivity is 1.

\Rightarrow if y is not used in evaluating y and y itself is used through a chain, then adaptivity is 1.

\Rightarrow if y is used in evaluating y and y itself is not used through a chain, then adaptivity is 1.

\Rightarrow if y is not used in evaluating y and y itself is not used through a chain, then adaptivity is 1.

\Rightarrow if y is used in evaluating y and y itself is used through a chain, then adaptivity is 1.

\Rightarrow if y is not used in evaluating y and y itself is not used through a chain, then adaptivity is 1.

\Rightarrow if y is used in evaluating y and y itself is not used through a chain, then adaptivity is 1.

\Rightarrow if y is not used in evaluating y and y itself is used through a chain, then adaptivity is 1.

\Rightarrow if y is not used in evaluating y and y itself is not used through a chain, then adaptivity is 1.

\Rightarrow if y is used in evaluating y and y itself is used through a chain, then adaptivity is 1.

\Rightarrow if y is not used in evaluating y and y itself is not used through a chain, then adaptivity is 1.

\Rightarrow if y is used in evaluating y and y itself is not used through a chain, then adaptivity is 1.

\Rightarrow if y is not used in evaluating y and y itself is used through a chain, then adaptivity is 1.

\Rightarrow if y is not used in evaluating y and y itself is not used through a chain, then adaptivity is 1.

\Rightarrow if y is used in evaluating y and y itself is not used through a chain, then adaptivity is 1.

\Rightarrow if y is not used in evaluating y and y itself is used through a chain, then adaptivity is 1.

\Rightarrow if y is not used in evaluating y and y itself is not used through a chain, then adaptivity is 1.

\Rightarrow if y is used in evaluating y and y itself is not used through a chain, then adaptivity is 1.

\Rightarrow if y is not used in evaluating y and y itself is used through a chain, then adaptivity is 1.

\Rightarrow if y is not used in evaluating y and y itself is not used through a chain, then adaptivity is 1.