

CS 5/6110, Software Correctness Analysis, Spring 2022

Ganesh Gopalakrishnan
School of Computing
University of Utah
Salt Lake City, UT 84112



Recap : This pertains to the quiz ... ask if I should show you how to get “Tau”...

Now consider the recursive definition:

$$F(x, y) = \text{if } x = y \text{ then } y + 1 \text{ else } F(x, F(x - 1, y + 1)).$$

$$f_1 = \lambda(x, y) . \text{if } x = y \text{ then } y + 1 \text{ else } x + 1$$

$$f_2 = \lambda(x, y) . \text{if } x \geq y \text{ then } x + 1 \text{ else } y - 1$$

$$f_3 = \lambda(x, y) . \text{if } x \geq y \text{ and } x - y \text{ is even then } x + 1 \text{ else } \perp$$

Function f_3 corresponds to $\lim_i \{ \text{Tau}^i [\text{Bottom_fn}] \}$

Where Tau for “F” above is: [...fill this...](#) and is called the “functional underlying the recursive definition (in Manna’s book)

In Chapter 18 of Book-3, it is called the “pre” function (e.g. PreFact etc) on which the Y combinator is applied.

Applying the Y combinator gives the same effect as computing the limit of this chain of functions

What does $\text{Tau}^1[\text{Bottom}]$ correspond to? What about Tau^2 ? Tau^3 ? [...fill this...](#)

What I'm trying to do

- Tell you that an elephant can be viewed from many sides
 - A tube
 - A pancake
 - A pokey thing
- Fixpoint theory is in many areas of CS
 - Context-free Languages
 - PL semantics
 - Static analysis
 - CTL model checking
 - Even CMOS transistor simulation
- Learning it in “just one class” may give you the view that elephants are pokey objects 😊

What fixed-point/fixpoint theory is aimed at

- Solving equations
 - Some equations make sense, some don't
 - $X = 2$ - yes
 - $X = X + 1$ - no, in int
 - $F(x) = F(x)$ - yes , but too many solutions
 - $F(x) = F(x+1)$ - <your answer>
 - $F(x) = F(x) + 1$ - <your answer_

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 - $F(x) = F(x) + 1$ - no
- Main points
 - Recursion / circularity is natural
 - How do we solve when we have circular situations?

Fixpoint Theory to Explain CFGs

- Context-free Grammars re-interpreted as recursive equations
 - $S \rightarrow aSbS \mid bSaS \mid SS \mid \text{epsilon}$
 - Versus
 - $S \rightarrow aSbS \mid bSaS \mid \text{epsilon}$
- How do we view the above as language equations?

Fixpoint Theory to Explain CFGs

- Context-free Grammars re-interpreted as recursive equations
 - $S \rightarrow aSbS \mid bSaS \mid \epsilon$
 - Solve the recursive language equation
- $L_S = \{a\} L_S \{b\} L_S \cup \{b\} L_S \{a\} L_S \cup \{\epsilon\}$

Fixpoint Theory to Explain CFGs

- Context-free Grammars re-interpreted as recursive equations
 - $S \rightarrow aSbS \mid bSaS \mid \text{epsilon}$
 - Solve the recursive language equation
- $L_S = \{a\} L_S \{b\} L_S \cup \{b\} L_S \{a\} L_S \cup \{e\}$
- What do we get when we iterate from $L_S = \{\}$ “upwards” ?

But for the grammar with the $| \dots |$ SS rule

- Context-free Grammars re-interpreted as recursive equations
 - $S \rightarrow aSbS \mid bSaS \mid SS \mid \text{epsilon}$
- $L_S = \{a\} L_S \{b\} L_S \cup \{b\} L_S \{a\} L_S \cup \{e\} \cup L_S L_S$
- We have two solutions!
 - What are they?

Fixpoint Theory to Explain CFGs

- Context-free Grammars re-interpreted as recursive equations
 - $S \rightarrow aSbS \mid bSaS \mid SS \mid \text{epsilon}$
- $L_S = \{a\} L_S \{b\} L_S \cup \{b\} L_S \{a\} L_S \cup \{e\} \cup L_S L_S$
- One is the “usual solution”
 - Context-free rewrite schemes (productions) go after the LFP
 - “equal a’s and b’s”
 - We also have Σ^* as a solution
 - For the case we have $S \rightarrow \dots \mid SS$

Uniqueness of Solutions

- People in general like things when solutions are unique
 - They sleep well at night
 - They are kinder to strangers, even smile at them
 - They remember to floss well at night
 - ...
- How about these situations : unique or not?
 - $X^2 = 4$
 - $X^2 = -4$
 - Quadratic equations
 - ...

Uniqueness of Solutions

- People in general like things when solutions are unique
- A lot of Fixpoint Theory in CS and programming is aimed at seeking uniqueness
- This is why equation systems on monotone lattices are important

Monotone is a “betterness order”

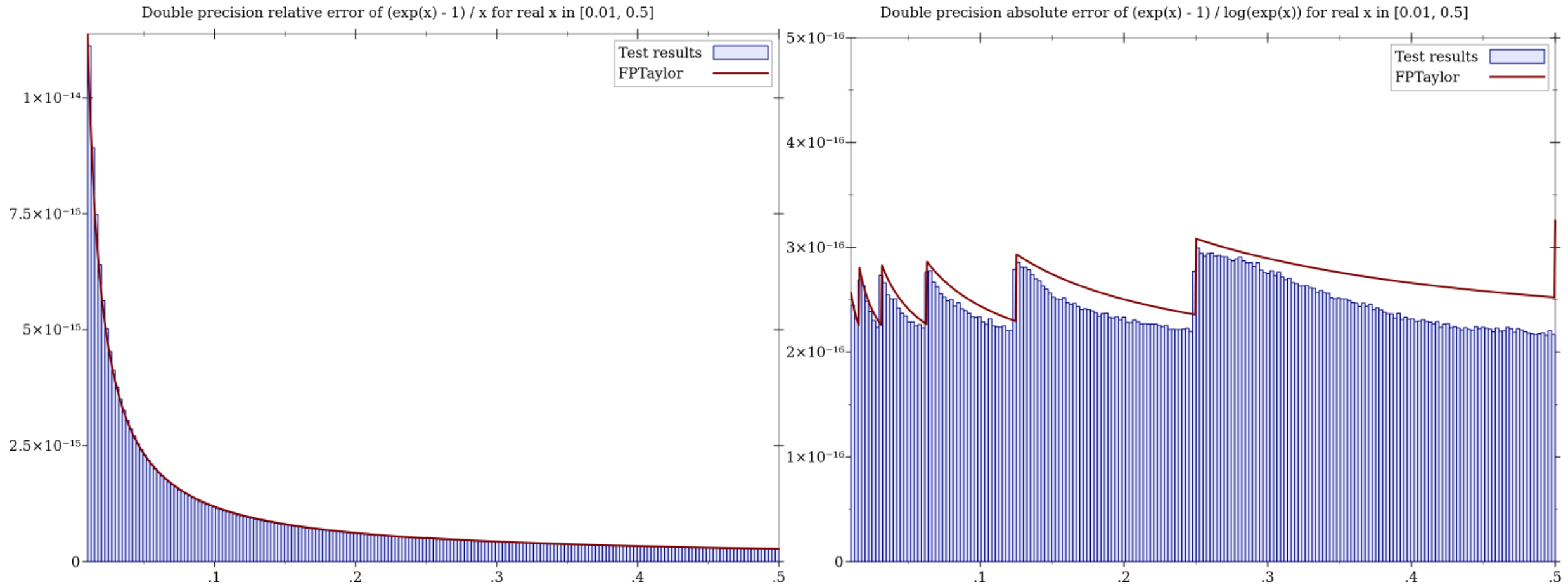
- Or a worseness order
 - $A \leq B$ means
 - A is a better component than B (in static analysis at least)
 - A is a tighter approximation than B
- Example
 - You can get resistors with 10% tolerance (a silver band on them)
 - Or a 5% tolerance (gold band on them)
 - $50\text{ohms}@5\% \leq 50\text{ohms}@10\%$
- Resistor parallel composition respects betterness
 - $R1 \parallel R2 = (R1 * R2) / (R1 + R2)$
 - Here if you initially use $R2 == R2@10\%$
 - And stick in $R2@5\%$
 - The whole ckt gets better
- This is monotonicity
 - Not all systems are monotonic
 - Hence causes huge debugging headaches when monotonicity is violated!

Floating-point error behaves non-monotone

- If you plug-in a component that introduces worse error, the overall error can decrease!
- See plots next slide!

Example of non-monotonicity (FP example)

- This example was discussed last class - here are the plots



Solving for a pair of unknowns is natural

- Mutual recursion, i.e. defining two languages
 - $S \rightarrow \text{epsilon} \mid (W S$
 - $W \rightarrow (W W \mid)$
- Solve
- $(L_S, L_W) = (\{e\} \cup \{ (L_W L_S \}, \{ (L_W L_W \cup \{ \} \})$

Fixpoint Theory to Explain CFGs

- Then discuss the system
 - $S \rightarrow \epsilon \mid (W S$
 - $W \rightarrow (W W \mid)$
- Solve
- $(L_S, L_W) = (\{e\} \cup \{ (L_W L_S \ , \ (L_W L_W \cup \{ \}) \})$
- What fixpoint obtained by iterating up from $(\{ \} , \{ \})$?
- What is the lattice ordering?

Uniqueness of Least Fixpoints

- Least fixpoints exist and are unique when τ is
 - Monotonic
 - Continuous
- For infinite lattices
 - Continuity implies Monotonicity
- For finite lattices
 - Monotonicity implies Continuity

Fixpoint Theory to Explain CFGs

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 - $S \rightarrow aSbS \mid bSaS \mid SS \mid \text{epsilon}$
 - Versus
 - $S \rightarrow aSbS \mid bSaS \mid \text{epsilon}$
- Then discuss the system
 - $S \rightarrow \text{epsilon} \mid (WS)$
 - $W \rightarrow (WW \mid)$

Fixpoint Theory to Explain CFGs

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 - $S \rightarrow aSbS \mid bSaS \mid \epsilon$
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Fixpoint Theory to Explain CFGs

- Context-free Grammars re-interpreted as recursive equations
 - $S \rightarrow aSbS \mid bSaS \mid SS \mid \text{epsilon}$
- $L_S = \{a\} L_S \{b\} L_S \cup \{b\} L_S \{a\} L_S \cup \{e\} \cup L_S L_S$
- Are there 2 fixpoints? Which is found using iteration using $\{\}$ as the bottom (going up)?

Fixpoint Theory to Explain CFGs

- Then discuss the system
 - $S \rightarrow \epsilon \mid (W S$
 - $W \rightarrow (W W \mid)$
- Solve
- $(L_S, L_W) = (\{e\} \cup \{() L_W L_S \} , \{() L_W L_W \cup \{\}\})$

Fixpoint Theory to Explain CFGs

- Then discuss the system
 - $S \rightarrow \epsilon \mid (W S$
 - $W \rightarrow (W W \mid)$
- Solve
- $(L_S, L_W) = (\{e\} \cup \{ (L_W L_S \}, \{ (L_W L_W \cup \{) \})$
- What fixpoint obtained by iterating up from $(\{\}, \{\})$?
- What is the lattice ordering?

Now consider the recursive definition:

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What does $\text{Tau}^1[\text{Bottom}]$ correspond to? What about Tau^2 ? Tau^3 ? [...fill this...](#)

Now discuss notes in this directory

- Manna's work on interpreting these functions
- Which function can we “experimentally compute”?
 - If we keep experimenting with F in say Python , what function table can we fill ?
 - What if we did it in a different (lazy) language)?
- That is, if we compute according to a fixpoint computation rule, we will get the “true answer”
- None of this is largely of concern for finite lattices
 - Many static-analysis situations
- But the general story is important to know.

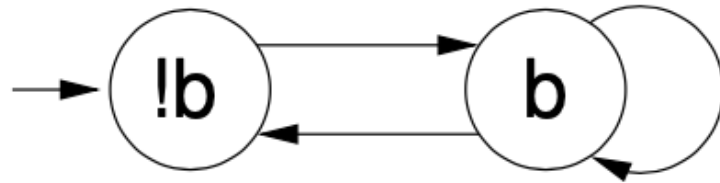
CTL Model Checking

- Least fixpoints exist and are unique when Tau is
 - Monotonic
 - Continuous
- For infinite lattices
 - Continuity implies Monotonicity
- For finite lattices
 - Monotonicity implies Continuity

State-Space Travel via BDDs

- We will use BDDs to represent Kripke Structure
- We will model Transition Relations using BDDs
- Use Boolean operations to obtain the set of reachable states

State Transition Systems via BDDs



$$\lambda(b, b').(b + b').$$

Fig. 11.4. Simple state transition system (example SimpleTR)

The values of b and b' for which this relation is satisfied represent the present and next states in our example. In other words,

- a move where b is false now and true in the next state is represented by $\neg b b'$.
- a move where b is true in the present and next states is represented by $b b'$.
- finally, a move where b is true in the present state and false in the next state is represented by $b \neg b'$.

The set of reachable states defined by “P”

In other words, we can introduce a predicate P such that a state x is in P if and only if it is reachable from the initial state I through a finite number of steps, as dictated by the transition relation T . The above recursive recipe is encoded as

$$P(s) = (I(s) \vee \exists x.(P(x) \wedge T(x, s))).$$

This can be computed via fixpoint iteration

Rewriting again, we have

$$P = (\lambda G.(\lambda s.(I(s) \vee \exists x.(G(x) \wedge T(x, s))))) P.$$

In other words, P is a fixed-point of

$$\lambda G.(\lambda s.(I(s) \vee \exists x.(G(x) \wedge T(x, s)))).$$

Let us call this Lambda expression H :

$$H = \lambda G.(\lambda s.(I(s) \vee \exists x.(G(x) \wedge T(x, s)))).$$

$$P_1 = \lambda G.(\lambda s.(I(s) \vee \exists x.(G(x) \wedge T(x, s))))P_0 \quad \text{Etc...}$$

This can be computed via fixpoint iteration

- $I = \lambda b. \neg b.$
- $T = \lambda(b, b'). (b + b').$
- $P_0 = \lambda s. false$, which encodes the fact that “we’ve reached nowhere yet!”
- $P_1 = \lambda G. (\lambda s. (I(s) \vee \exists x. (G(x) \wedge T(x, s)))) P_0.$
This simplifies to $P_1 = I$, which is, in effect, an assertion that we’ve “just reached” the initial state, starting from P_0 .
- Let’s see the derivation of P_1 in detail. Expanding T and P_0 , we have
$$P_1 = \lambda G. (\lambda s. (I(s) \vee \exists x. (G(x) \wedge (x + s)))) (\lambda x. false).$$

This can be computed via fixpoint iteration

- The above simplifies to $\neg b$.
- By this token, we are expecting P_2 to be all states that are zero or one step away from the start state. Let's see whether we obtain this result.
- $$\begin{aligned} P_2 &= \lambda G.(\lambda s.(I(s) \vee \exists x.(G(x) \wedge T(x, s))))P_1. \\ &= \lambda s.(\neg s \vee \exists x.(\neg x \wedge (x + s))). \\ &= \lambda s.1. \end{aligned}$$

Forward Reachability via the BDD tool called “BED”

```

var b bp           % Declare b and b'
let I = !b         % Declare init state
let t1 = !b and bp % 0 --> 1
upall t1           % Build BDD for it
view t1            % View it
let t2 = b and bp  % 1 --> 1
let t3 = b and !bp % 1 --> 0
let T = t1 or t2 or t3 % All three edges
upall T            % Build and view the BDD
view T             %

```

```

let P0 = false
upall P0
view P0

```

```

let P1 = I or ((exists b. (P0 and T))[bp:=b])
upall P1
view P1

```

```

let P2 = I or ((exists b. (P0 and T))[bp:=b])
upall P2
view P2

```

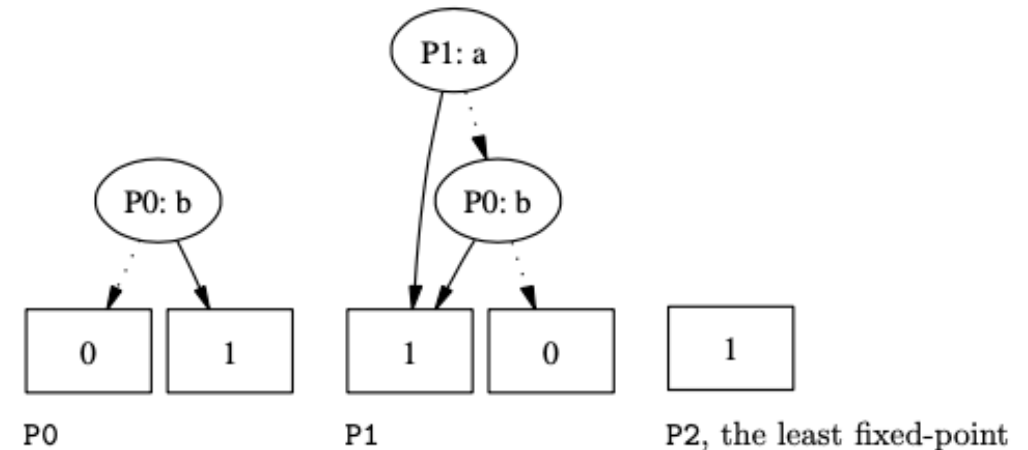


Fig. 11.5. BED commands for reachability analysis on SimpleTR, and the fixed-point iteration leading up to the least fixed-point that denotes the set of reachable states starting from I

Forward Reahability
via the BDD tool
called “BED”:
another example

```
var a ap b bp
```

```
let T = (a and b and ap and bp) or /* S0 -> S0 */
        (!a and b and !ap and bp) or /* S1 -> S1 */
        (a and !b and ap and !bp) or /* S2 -> S2 */
        (!a and !b and !ap and !bp) or /* S3 -> S3 */
        (!a and b and ap and !bp) or /* S1 -> S2 */
        (a and !b and !ap and bp) or /* S2 -> S1 */
        (!a and b and ap and bp) or /* S1 -> S0 */
        (a and !b and ap and bp) /* S2 -> S0 */
```

```
upall T
```

```
view T /* Produces BDD for TREL 'T' */
```

```
let I = a and b
```

```
let P0 = b
```

```
let P1 = I or ((exists a. (exists b. (P0 and T)))[ap:=a][bp:=b])
```

```
upall P1
```

```
view P1
```

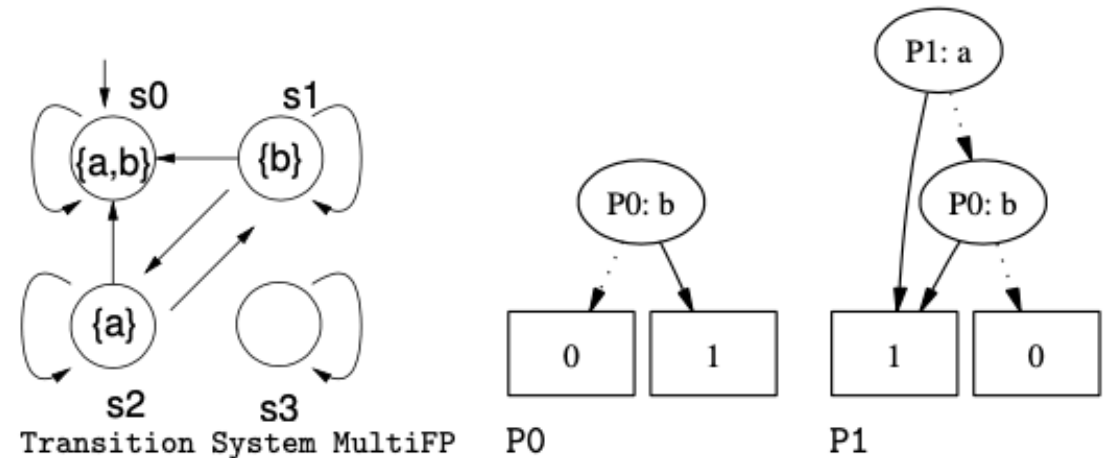


Fig. 11.6. Example where multiple fixed-points exist. This figure shows attainment of a fixed-point $a \vee b$ which is between the least fixed-point of $a \wedge b$ and the greatest fixed-point of 1. The figure shows the initial approximant P0 and the next approximant P1

Now, for the CTL
Logic, and
Reachability using
BDDs

CTL formulas are Kripke structure classifiers

Given a CTL formula φ , all possible computation trees fall into two bins—*models* and *non-models*.⁵ The computation trees in the *model* ('good') bin are those that satisfy φ while those in the *non-model* ('bad') bin obviously falsify φ .

Consider the CTL formula $\text{AG} (\text{EF} (\text{EG } a))$ as an example. Here,

- 'A' is a *path quantifier* and stands for *all paths* at a state
- 'G' is a *state quantifier* and stands for *everywhere along the path*
- 'E' is a *path quantifier* and stands for *exists a path*
- 'F' is a *state quantifier* and stands for *find (or future)* along a path
- 'X' is a *state quantifier* and stands for *next* along a path

The truth of the formula $\text{AG} (\text{EF} (\text{EG } a))$ can be calculated as follows:

- In all paths, everywhere along those paths, $\text{EF} (\text{EG } a)$ is true
- The truth of $\text{EF} (\text{EG } a)$ can be calculated as follows:
 - There exists a path where we will find that $\text{EG } a$ is true.
 - The truth of $\text{EG } a$ can be calculated as follows:
 - * There exists a path where a is globally true.

Now, for the CTL
Logic, and
Reachability using
BDDs

CTL formulas γ are inductively defined as follows:

$\gamma \rightarrow x$	a propositional variable	
$\neg\gamma$	negation of γ	
(γ)	parenthesization of γ	
$\gamma_1 \vee \gamma_2$	disjunction	
$\text{AG } \gamma$	on all paths,	everywhere along each path
$\text{AF } \gamma$	on all paths,	somewhere on each path
$\text{AX } \gamma$	on all paths,	next time on each path
$\text{EG } \gamma$	on some path,	everywhere on that path
$\text{EF } \gamma$	on some path,	somewhere on that path
$\text{EX } \gamma$	on some path,	next time on that path
$\text{A}[\gamma_1 \text{ U } \gamma_2]$	on all paths,	γ_1 until γ_2
$\text{E}[\gamma_1 \text{ U } \gamma_2]$	on some path,	γ_1 until γ_2
$\text{A}[\gamma_1 \text{ W } \gamma_2]$	on all paths,	γ_1 weak-until γ_2
$\text{E}[\gamma_1 \text{ W } \gamma_2]$	on some path,	γ_1 weak-until γ_2

Now, for the CTL
Logic, and
Reachability using
BDDs

$$EG\ p = p \wedge (EX\ (EG\ p))$$

```
bed> var a a1 b b1
```

```
var a a1 b b1
```

```
bed> let TREL =
```

```
    (not(a) and b and a1 and not(b1)) or (a and not(a1) and b1) or
```

```
    (a and not(b) and b1)                or (a and not(b) and a1)
```

```
bed> upall TREL
```

```
Upall( TREL ) -> 53
```

```
bed> view TREL ... (displays the BDD)
```


Now, for the CTL
Logic, and
Reachability using
BDDs

$$EG\ p = p \wedge (EX\ (EG\ p))$$

- In the BED syntax, $a \oplus b$ is written `a != b`. Now we perform the fixed-point iteration assisted by BED. We construct variable names that mnemonically capture what we are achieving at each step:

```
EG_a_xor_b_0 = true -- first approximant
```

```
EG_a_xor_b_1 = (a != b) and (EX true) -- second approximant
```

This simplifies to `(a != b)`, as `(EX true)` is `true`.

Now, in order to determine `EG_a_xor_b_2`, we continue the fixed-point iteration process, and write

```
EG_a_xor_b_2 = (a != b) and EX (a != b)
```

At this juncture, we realize that we need to calculate `EX (a != b)`. This can be calculated using BED as follows:

```
bed> let EX_a_xor_b = exists a1. exists b1. (TREL and (a1 != b1))
bed> upall EX_a_xor_b
bed> view EX_a_xor_b
```

Now, for the CTL
Logic, and
Reachability using
BDDs

$$EG\ p = p \wedge (EX\ (EG\ p))$$

Calculating AX

If we have to calculate $AX\ p$, we would employ duality and write it as

$$\neg (EX\ \neg p)$$

This approach will be used in the rest of this book.

Now, for the CTL
Logic, and
Reachability using
BDDs

$$A[pUq] = q \vee (p \wedge AX(A[pUq]))$$

```
bed> var p p1 q q1
bed> let TREL = (p and not(q) and p1 and not(q1))
               or (p and not(q) and p1 and q1)
               or (p and q and not(p1) and q1)
               or (not(p) and q and p1 and not(q1))
               or (p and not(q) and not(p1) and q1)
               or (p and not(q) and p1 and not(q1))

bed> upall TREL
Upall( TREL ) -> 67
bed> view TREL
bed> let A_p_U_q_0 = false
bed> let AX_A_p_U_q_0 = false
bed> let A_p_U_q_1 = (q or (p and AX_A_p_U_q_0))
```

Now, for the CTL
Logic, and
Reachability using
BDDs

$$A[pUq] = q \vee (p \wedge AX(A[pUq]))$$

```
bed> upall A_p_U_q_1
Upall( A_p_U_q_1 ) -> 3
bed> view A_p_U_q_1
bed> let EX_not_q = exists p1. exists q1. (TREL and !q1)
bed> upall EX_not_q
Upall( EX_not_q ) -> 80
bed> view EX_not_q
bed> let AX_q = !EX_not_q
bed> upall AX_q
Upall( AX_q ) -> 82
bed> view AX_q
bed> let A_p_U_q_2 = (q or (p and AX_q))
bed> upall A_p_U_q_2
Upall( A_p_U_q_2 ) -> 3
bed> view A_p_U_q_2 --> gives ‘‘q’’, hence denotes {S1,S2} -- LFP
```

Now, for the CTL
Logic, and
Reachability using
BDDs

$$A[pUq] = q \vee (p \wedge AX(A[pUq]))$$

23.2.5 GFP for Until

```
bed> let A_p_U_q_0 = true
bed> let AX_A_p_U_q_0 = true
bed> let A_p_U_q_1 = (q or (p and AX_A_p_U_q_0))
bed> upall A_p_U_q_1
Upall( A_p_U_q_1 ) -> 72
view A_p_U_q_1
bed> let EX_not_p_or_q = exists p1. exists q1. (TREL and !(p1 or q1))
bed> upall EX_not_p_or_q
Upall( EX_not_p_or_q ) -> 0
bed> let AX_p_or_q = !EX_not_p_or_q
bed> upall AX_p_or_q
Upall( AX_p_or_q ) -> 1
bed> view A_p_U_q_1
bed> let A_p_U_q_2 = (q or (p and AX_p_or_q)) --> reached
Fixed-point (q or p) which denotes {S0,S1,S2,S3}
```

Summary

- Fixpoint theory is everywhere in CS
 - Static analysis
 - Recursive program analysis
 - CFG explanation
 - CTL model-checking
- Finding lattices and monotonic + continuous functionals is key
- Once set up this way, we usually go after the least fixpoint
- Greatest fixpoints also “make sense”
 - But sometimes they are useless
 - as in the CFG example $S \rightarrow aSbS \mid bSaS \mid SS \mid \epsilon$

Summary