

Least and Greatest Fixpoints in Second-Order Logic

Giuseppe De Giacomo

Construction principle

To define a set Z , here denoted by a predicate $Z(\vec{x})$, we need to say what its elements are.

Construction Principle

The *construction principle* tells us how to obtain these elements recursively.

$$\forall \vec{x}. Z(\vec{x}) \equiv \Phi(Z, \vec{x}) \quad (1)$$

In this case Φ is called a *constructor* for Z .

Any solution of this recursive equation is called a *fixpoint* of the operator Φ .

To define a set Z , here denoted by a predicate $Z(\vec{x})$, we need to say what its elements are, that is we need to choose a solution.

Example: natural numbers

Example (Natural Numbers)

We often define Natural Number as:

- $0 \in \mathbb{N}$
- $\forall x. x \in \mathbb{N} \supset succ(x) \in \mathbb{N}$
- nothing else in \mathbb{N}

If we want to fully formalize this definition, we can start by defining the corresponding *construction principle* $\forall \vec{x}. N(\vec{x}) \equiv \Phi(N, \vec{x})$, which in this case is:

$$\forall x. (N(x) \equiv x = 0 \vee \exists y. x = succ(y) \wedge N(y))$$

The *constructor*, $\Phi(N, \vec{x})$, is

$$x = 0 \vee \exists y. x = succ(y) \wedge N(y)$$

Formally $\mathbb{N}(x)$ is the least fixpoint of such constructor, i.e., the smallest solution that satisfy this construction principle.

Least fixpoint

The Tarski-Knaster Theorem guarantees that if the operator Φ is monotone, the equation (1) has both a least and a greatest solution.

A sufficient condition for monotonicity is that all occurrences of Z occur within a even number of negations. This condition must be always satisfied.

Least Fixpoint in Second-Order Logic

Specifically the smallest solution of the equation (1), i.e., the *least fixpoint* is characterized in Second-Order Logic (SOL) as:

$$\begin{aligned} Z^\mu(\vec{x}) \equiv \quad & \forall P. (\\ & \quad \forall \vec{y}. (\Phi(P, \vec{y}) \supset P(\vec{y})) \\ & \quad \supset P(\vec{x})) \end{aligned} \tag{2}$$

NB: (2) is indeed a re-writing in SOL of Tarski-Knaster's characterization of the least fixpoint:

$$Z^\mu = \bigcap \{ \mathcal{P} \mid \Phi(\mathcal{P}) \subseteq \mathcal{P} \}$$

Example: natural numbers

Example (Natural Numbers)

Formally $\mathbb{N}(x)$ is the least fixpoint of such constructor, i.e., the smallest solution that satisfy the construction principle.

Namely considering that the constructor $\Phi(N, \vec{x})$ is

$$x = 0 \vee \exists y. x = \text{succ}(y) \wedge N(y)$$

We can define Natural Numbers $\mathbb{N}(x)$ in Second Order Logic as:

$$\begin{aligned} \mathbb{N}(x) \equiv \quad & \forall N. (\\ & \forall x. ((x = 0 \vee \exists y. x = \text{succ}(y) \wedge N(y)) \supset N(x)) \\ & \supset N(x)) \end{aligned} \tag{3}$$

Note N in the constructor above is not in the scope of any negation, hence the constructor is a (syntactically) monotonic operator, and Tarski-Knaster fixpoint theorem applies: a unique least fixpoint exists.

Induction principle

Let Z^μ be the smallest solution to (1). Then the *induction principle* holds.

Induction Principle

$$\begin{aligned} \forall P, \vec{x}. (& \\ & \forall \vec{y}. (\Phi(P, \vec{y}) \supset P(\vec{y})) \\ & \supset (Z^\mu(\vec{x}) \supset P(\vec{x}))) \end{aligned} \tag{4}$$

i.e., whatever P satisfying the the recursive specification, we have that Z^μ is included in it.

A set Z satisfying construction principle (1) and induction principle (4) is indeed the smallest solution Z^μ to (1), i.e., *least fixpoint* of an operator $\Phi(P, \vec{y})$.

Induction principle: least fixpoint

We can rewrite the induction principle (4) in the following way

$$\begin{aligned} \forall \vec{x}. (Z^\mu(\vec{x}) \supset \forall P. (& \\ & \forall \vec{y}. (\Phi(P, \vec{y}) \supset P(\vec{y})) \\ & \supset P(\vec{x}))) \end{aligned} \tag{5}$$

Notice that implication in the opposite direction follows from the construction principle (1).

We obtain

$$\begin{aligned} Z^\mu(\vec{x}) \equiv \forall P. (& \\ & \forall \vec{y}. (\Phi(P, \vec{y}) \supset P(\vec{y})) \\ & \supset P(\vec{x})) \end{aligned} \tag{6}$$

which is formal definition of a least fixpoint!

Example: natural numbers

Example (Natural Numbers)

The induction principle for the Natural Numbers is:

$$\begin{aligned} \forall P. (& \hspace{15em} (7) \\ & \forall x. ((x = 0 \vee \exists y. x = succ(y) \wedge P(y)) \supset P(x)) \\ & \supset (\mathbb{N}(x) \supset P(x))) \end{aligned}$$

This formula can be rewritten in a more familia format:

$$\begin{aligned} \forall P. (& \hspace{15em} (8) \\ & \forall x. (\\ & \quad (x = 0 \supset P(x)) \hspace{5em} \textit{base case} \\ & \quad \wedge \\ & \quad (\forall y. x = succ(y) \wedge P(y) \supset P(x)) \hspace{2em} \textit{inductive case} \\ & \supset (\mathbb{N}(x) \supset P(x))) \end{aligned}$$

This is the formula that we implicitly use when we do prove by induction.

Greatest fixpoint

Consider again the construction principle:

$$\forall \vec{x}. Z(\vec{x}) \equiv \Phi(Z, \vec{x}) \quad (9)$$

Greatest Fixpoint in Second-Order Logic

The greatest solution of the equation (1), i.e., the greatest fixpoint is characterized in Second-Order Logic (SOL) as:

$$\begin{aligned} Z^\nu(\vec{x}) \equiv \exists P. (& \\ & \forall \vec{y}. P(\vec{y}) \supset \Phi(P, \vec{y})) \\ & \wedge P(\vec{x}) \end{aligned} \quad (10)$$

NB: (10) is indeed a re-writing in SOL of Tarski-Knaster's characterization of the greatest fixpoint:

$$Z^\nu = \bigcup \{ \mathcal{P} \mid \mathcal{P} \subseteq \Phi(\mathcal{P}) \}$$

Coinduction principle: greatest fixpoint

Let Z^ν is the largest solution of (1). Then the *coinduction principle* holds.

Coinduction Principle

$$\begin{aligned} \forall P, \vec{x}. (& \\ & \forall \vec{y}. (P(\vec{y}) \supset \Phi(P, \vec{y})) \\ & \supset (P(\vec{x}) \supset Z^\nu(\vec{x}))) \end{aligned} \quad (11)$$

i.e., whatever solution P of the recursive specification we take, Z^ν includes it.

Note that, we can rewrite the coinduction principle (11) in the following way

$$\begin{aligned} \forall \vec{x}. (& \\ & (\exists P. (\forall \vec{y}. P(\vec{y}) \supset \Phi(P, \vec{y})) \wedge P(\vec{x})) \\ & \supset Z^\nu(\vec{x})) \end{aligned} \quad (12)$$

Notice that implication in the opposite direction follows from the construction principle (1). Combining the two, we get back the definition of greatest fixpoint 10.

Example: bisimulation

Example (Bisimulation)

We define bisimulation as:

- $(s, t) \in R \supset \Pi_1(s) = \Pi_2(t) \wedge$
 $\forall a. \delta_1(s, a, s') \supset \exists t'. \delta_2(t, a, t') \wedge (s', t') \in R \wedge$
 $\forall a. \delta_2(t, a, t') \supset \exists s'. \delta_1(s, a, s') \wedge (s', t') \in R$

The corresponding *construction principle*:

$$\begin{aligned} R(s, t) \equiv & (\Pi_1(s) = \Pi_2(t)) \wedge \\ & (\forall a. \delta_1(s, a, s') \supset \exists t'. \delta_2(t, a, t') \wedge R(s', t')) \wedge \\ & (\forall a. \delta_2(t, a, t') \supset \exists s'. \delta_1(s, a, s') \wedge R(s', t')) \end{aligned}$$

Example: bisimulation

Example (Bisimilarity - Greatest Fixpoint)

Bisimilarity, denoted by \sim is the largest bisimulation, i.e., the largest solution of the construction principle above, i.e., the greatest fixpoint. In second-order logic:

$$\begin{aligned}(\hat{s} \sim \hat{t}) \equiv & \\ & \exists R. (\forall s, t. R(s, t) \supset (\Pi_1(s) = \Pi_2(t)) \wedge \\ & \quad (\forall a. \delta_1(s, a, s') \supset \exists t'. \delta_2(t, a, t') \wedge R(s', t')) \wedge \\ & \quad (\forall a. \delta_2(t, a, t') \supset \exists s'. \delta_1(s, a, s') \wedge R(s', t'))) \\ & \wedge \\ & R(\hat{s}, \hat{t})\end{aligned}$$

Example: bisimulation

Example (Coinduction Principle for Bisimilarity)

Bisimilarity, denoted by \sim is the largest bisimulation, i.e., the largest solution of the construction principle above, i.e., the greatest fixpoint. In second-order logic:

$\forall R, \hat{s}, \hat{t}.$

$$\begin{aligned} (\forall s, t. R(s, t) \supset & (\Pi_1(s) = \Pi_2(t)) \wedge \\ & (\forall a. \delta_1(s, a, s') \supset \exists t'. \delta_2(t, a, t') \wedge R(s', t')) \wedge \\ & (\forall a. \delta_2(t, a, t') \supset \exists s'. \delta_1(s, a, s') \wedge R(s', t'))) \end{aligned}$$

i.e., R is a bisimulation

$$\supset (R(\hat{s}, \hat{t}) \supset (\hat{s} \sim \hat{t}))$$