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Probabilistic Programming
Exercise Sheet 4 Solutions
Group 7

P.S. Solutions start from next page!

¹
(a) Given, a Markov Chain $D = (\Sigma, \tau_i, P)$ with goal set $G \subseteq \Sigma$.
Define $\Sigma_? = \Sigma \setminus G$, matrix $A = (a_{ij}) = (P(\tau_i, \tau_j))_{\tau_i, \tau_j \in \Sigma_?}$, and vector $b = (b_\tau)_{\tau \in \Sigma_?}$, where $b_\tau = \sum_{\sigma \in G} P(\tau, \sigma)$.

We define a partial order on the vectors in V , where these vectors have entries indexed by the states in $\Sigma_?$ with functions of type $\Sigma_? \rightarrow [0, 1]$. The partial order is as follows—

$(V, \text{"}\leq\text{"})$, where $\underline{v}_1 \leq \underline{v}_2 \Rightarrow$ every element of $\underline{v}_1 \leq$ every element of \underline{v}_2 ;

Note that, every vector in V is of order $|\Sigma_?| \times 1$.

Proof that $(V, \text{"}\leq\text{"})$ is a partial order :-

- For all $\underline{x} \in V$, $\underline{x} = \underline{x}$; hence reflexive
- For any $\underline{x}_1, \underline{x}_2, \underline{x}_3 \in V$,

$$x_{1i} \leq x_{2i} \quad \forall i=1(1)|\Sigma_?| \\ \text{and, } x_{2i} \leq x_{3i} \quad \forall i=1(1)|\Sigma_?| \Rightarrow x_{1i} \leq x_{3i} \quad \forall i=1(1)|\Sigma_?|$$

$\therefore \underline{x}_1 \leq \underline{x}_2$ and $\underline{x}_2 \leq \underline{x}_3 \Rightarrow \underline{x}_1 \leq \underline{x}_3$; hence transitive

- For $\underline{x}_1, \underline{x}_2 \in V$,

if $x_{1i} \leq x_{2i} \quad \forall i=1(1)|\Sigma_?|$ and $x_{2i} \leq x_{1i} \quad \forall i=1(1)|\Sigma_?|$
then $x_{1i} = x_{2i} \quad \forall i=1(1)|\Sigma_?|$

i.e., $\underline{x}_1 \leq \underline{x}_2$ and $\underline{x}_2 \leq \underline{x}_1 \Rightarrow \underline{x}_1 = \underline{x}_2$; hence anti-symmetric

Therefore, $(V, \text{"}\leq\text{"})$ is a partial order.

To prove :-

$f : V \rightarrow V$; $x \mapsto Ax + b$ has a least fixed point in this order.

By Kleene's fixed point theorem, f has a lfp if -

- (V, \sqsubseteq) is a complete lattice
- f is Scott continuous

Complete lattice -

Here, V is a set of vectors where the vectors have dimension $|S| \times 1$ and entries in $[0, 1]$. Therefore, there exists a well defined order between every pair of elements under (V, \sqsubseteq) . Since D is a finite state MC, therefore every subset S has a supremum and infimum in D wrt the relation " \sqsubseteq ".

Hence, (V, \sqsubseteq) is a complete lattice.

Scott continuity -

We have $f: V \rightarrow V : x \mapsto Ax + b$, where A and b are as defined above.

Let, $x_0 = \text{LS}$, where $S \subseteq V$ is a non-empty chain.

Then, $f(x_0) = Ax_0 + b \sqsubseteq Ax_0 + b \quad \forall x_0 \neq x_0 \text{ and } x \in V$, since A is positive semi-definite and every element of b is non-negative.

Therefore, order of elements in S are preserved under f . In fact,

S is clearly finite. Hence -

$$f(\text{LS}) = \text{L} f(S) \Rightarrow f \text{ is Scott-continuous}$$

\therefore By Kleene's fixpoint theorem, f has lfp wrt. this order (V, \sqsubseteq) . \blacksquare

(b) By Kleene's fixed point theorem, we can approximate the lfp of f by iteratively applying f as follows -

0 : Apply f to $\perp = \underline{\underline{0}}$

1 : Calculate $f(\perp) = b$

2 : Calculate $f(f(\perp)) = f(b) = Ab + b$

3 : Calculate $f^3(\perp), f^4(\perp)$ and so on

4 : For a large enough $n \in \mathbb{N}$, we should be able to reach $\text{lfp } f := \bigcup_{n \in \mathbb{N}} f^n(\perp)$

In fact, we essentially calculate -

$$f^n(\perp) = A^{n-1}b + A^{n-2}b + \dots + Ab + b$$

and take $\bigcup_{n \in \mathbb{N}} f^n(\perp)$.

2) a) $P: \text{if } (x > 0) \{ x := y + 1 \} \text{ else } \{ \text{if } (x = 0) \{ \text{skip} \} \text{ else } \{ x := x + 1 \} \}$

using Backward Reasoning:

$$WP \left[\left[\text{if } (x > 0) \{ x := y + 1 \} \text{ else } \{ \text{if } (x = 0) \{ \text{skip} \} \text{ else } \{ x := x + 1 \} \} \right] (x > y) \right]$$

$$= ((x > 0) \wedge WP[x := y + 1](x > y)) \vee ((x \leq 0) \wedge WP[\text{if } (x = 0) \{ \text{skip} \} \text{ else } \{ x := x + 1 \}](x > y))$$

$$= \underbrace{\left((x > 0) \wedge \underbrace{(y + 1 > y)}_{\text{true}} \right)}_{(x > 0)} \vee \left((x \leq 0) \wedge \left((x = 0) \wedge WP[\text{skip}](x > y) \right) \vee \left((x \neq 0) \wedge WP[x := x + 1](x > y) \right) \right)$$

$$= (x > 0) \vee ((x \leq 0) \wedge (x = 0) \wedge (x > y)) \vee ((x \neq 0) \wedge (y + 1 > y))$$

$$= (0 > y) \vee ((x = 0) \wedge (y + 1 > y))$$

b) $P: \text{while } (x \geq 1) \{ a := a * x; x := x - 1 \}$
 $f: a = x!$

The loop characteristic function $\phi_F(f)$ is:

$$\left((x \geq 1) \wedge WP[a := a * x; x := x - 1](X) \right) \vee \left((x < 1) \wedge (a = x!) \right)$$

We've seen in the lecture that $\phi_F: D \rightarrow \mathbb{P}$ is Scott continuous.
Using Kleene's fixed point theorem we get:

$$\text{lfp } \phi_F = \sup_{n \in \omega} \phi_F^n(\text{False})$$

↑
↓ in (P, \Rightarrow)

$$\Rightarrow \Phi_f(\text{False}) = ((x \geq 1) \wedge \underbrace{\text{WP}[\alpha := \alpha * x ; x := x - 1]}_{\text{false}}(\text{false})) \vee ((x < 1) \wedge (\alpha = x_!))$$

$$= (x < 1) \wedge (\alpha = x_!)$$

$$\Phi_f^2(\text{false}) = \Phi_f((x < 1) \wedge (\alpha = x_!)) =$$

$$= ((x \geq 1) \wedge \text{WP}[\alpha := \alpha * x ; x := x - 1] / ((x < 1) \wedge (\alpha = x_!))) \vee ((x < 1) \wedge (\alpha = x_!))$$

$$= ((x \geq 1) \wedge \text{WP}[\alpha := \alpha * x / (x < 2) \wedge (\alpha = (x - 1)_!)]) \vee ((x < 1) \wedge (\alpha = x_!))$$

$$= ((x \geq 1) \wedge ((x < 2) \wedge (\alpha = (x - 1)_!))) \vee ((x < 1) \wedge (\alpha = x_!))$$

$$= ((x = 1) \wedge (\alpha = (x - 1)_!)) \vee ((x < 1) \wedge (\alpha = x_!))$$

$$= (\alpha = 1) \vee ((x < 1) \wedge (\alpha = x_!))$$

$$\Phi^3(\text{false}) = \Phi_f((\alpha = 1) \vee ((x < 1) \wedge (\alpha = x_!))) =$$

$$= ((x \geq 1) \wedge \text{WP}[\alpha := \alpha * x ; x := x - 1] / (\alpha = 1) \vee ((x < 1) \wedge (\alpha = x_!))) \vee ((x < 1) \wedge (\alpha = x_!))$$

$$= ((x \geq 1) \wedge \text{WP}[\alpha := \alpha * x / (\alpha = 1) \vee ((x < 2) \wedge (\alpha = (x - 1)_!))]) \vee ((x < 1) \wedge (\alpha = x_!))$$

$$= ((x \geq 1) \wedge ((\alpha = 1) \vee ((x < 2) \wedge (\alpha = (x - 1)_!)))) \vee ((x < 1) \wedge (\alpha = x_!))$$

$$= ((\alpha = 1) \vee ((x < 1) \wedge (\alpha = x_!)))$$

\Rightarrow Thus, by computing the loop-characterising functional $\Phi_f(f)$, we found, using Kleene's fixed point theorem that $w\varphi(p, f)$ is in fact $(a=1) \vee ((x < 1) \wedge (a = x!))$

③ Monotonicity of weakest pre-expectations

@ (D, \sqsubseteq) - a complete lattice and $f, g : D \rightarrow D$ monotonic such that $\forall d \in D, f(d) \sqsubseteq g(d)$.

Using extended Knaster-Tarski Theorem, we can directly write $\text{lfp } f = \bigcap S$, where $S = \{d \in D \mid \underbrace{f(d)}_{\sqsubseteq} d\}$.

Also, $\text{lfp } g = \bigcap S'$, where $S' = \{d \in D \mid g(d) \sqsubseteq d\}$

Consider any such S' as above. $\forall d \in S' \subseteq D$, $f(d) \sqsubseteq g(d) \sqsubseteq d$. This means all elements d in any set S' also satisfies $\{d \in D \mid f(d) \sqsubseteq d\}$.

So, we have more sets $S = \{d \in D \mid f(d) \sqsubseteq d\}$ than sets S' satisfying $S' = \{d \in D \mid g(d) \sqsubseteq d\}$.

Now, since $\bigcap S$ or $\bigcap S'$ represents intersections of such states and we have more sets like S than S' , by basic properties of intersection

$$\bigcap S \sqsubseteq \bigcap S' \text{ which gives us}$$

$$\text{lfp } f = \bigcap S \sqsubseteq \bigcap S' = \text{lfp } g$$

$$\Rightarrow \text{lfp } f \sqsubseteq \text{lfp } g$$

□

(b) For any GCL program P , the function $\text{wp}(P) : \mathbb{P} \rightarrow \mathbb{P}$ takes a predicate as post condition and returns the predicate which is the weakest precondition of P w.r.t to the given post condition.

We consider the ordering $\sqsubseteq = \Rightarrow$ on \mathbb{P} .

So, we have $F \sqsubseteq G$ iff $F \Rightarrow G$.

Now, $\text{wp}[P](F)$ refers to the precondition that reaches/terminates in a state $s \models F$ and hence $\text{wp}[P](F) \Rightarrow F$

By transitivity of \Rightarrow

$$\text{wp}[P](F) \Rightarrow G \quad \dots \quad (1)$$

Similarly, $\text{wp}[P](G)$ refers to the pre-condition that reaches a state $s \models G$, which gives us $\text{wp}[P](G) \Rightarrow G \quad \dots \quad (2)$

So if $H = \text{wp}[P](G)$, then P starts in state $r \models H$ and terminates in a state $s \models G$.

Now for the post condition G , H can be wp iff. P starts in a state in H and terminates in G . Considering that P can also start from a state in $wp[P](F)$ and terminate at G (by ①), either H and $wp[P](F)$ are same or

$wp[P](F)$ can reach H , meaning

$$wp[P](F) \Rightarrow H = wp[P](G)$$

which proves that $wp(P): P \rightarrow P$ is monotonic.



④ Weakest liberal pre-conditions

① Without using the diverge statement a GCL program P along with a post-condition f can be such that $\text{wp}(P, f) \neq \text{wlp}(P, f)$ if we have a loop in the program P as
 $\text{wlp}(P, f) = \text{wp}[\text{P}](f) = \text{wp}[\text{P}](f)$ by construction.

Consider the following program:

P : while ($x \neq 0$) { $x := x - 1$ } and

post-condition

f : $x = 0$.

$$\begin{aligned} \text{We can see, to calculate } \text{wp}(P, f), \phi(x) &= (x=0 \wedge x=0) \vee_{x \neq 0} \\ &\quad \text{wp}[x := x - 1](x) \\ &= (x=0) \vee (x \neq 0 \wedge \text{wp}[x := x - 1](x)) \end{aligned}$$

$$\begin{aligned} \phi^0(\text{false}) &\Rightarrow \text{false}, \phi^1(\text{false}) = (x=0) \vee (x \neq 0 \wedge \text{wp}[x := x - 1](\text{false})) \\ &= (x=0) \vee (x \neq 0 \wedge \text{false}) \\ &\Rightarrow (x=0) \vee \text{false} = (x=0) \end{aligned}$$

$$\begin{aligned} \phi^2(\text{false}) &= (x=0) \vee (x \neq 0 \wedge \text{wp}[x := x - 1](x=0)) \\ &\Rightarrow (x=0) \vee (x \neq 0 \wedge x - 1 = 0) \\ &= (x=0) \vee (x=1) \end{aligned}$$

$$\phi^n(\text{false}) = (x=0) \vee (x=1) \dots \vee (x=n)$$

$$\Rightarrow \sup_{n \in \mathbb{N}} \phi^n(\text{false}) = \bigvee_{k=0}^n (x=k) = x \geq 0, \text{ or } x \in \mathbb{N}$$

So, starting with $x \geq 0$ terminates with $x=0$, starting with $x < 0$ doesn't terminate.

while $\text{wlp}(P, f) = \text{true}$

$\therefore \text{wp}(P, f) \neq \text{wlp}(P, f)$.

(b) P: while ($x \neq 0$) $\{x := x - 2\}$

$$\varphi(x) = (x=0 \wedge f) \vee (x \neq 0 \wedge \text{wfp}[\boxed{x := x-2}](x))$$

loop characteristics fn $\psi_f(x)$

$$\text{wfp}[\boxed{P}](f) = \text{gfp } x. \psi_f(x).$$

We know from construction of wfp and Kleene's fixed-point theorem that $\text{gfp } \psi_f = \inf_{n \in \mathbb{N}} \psi_f^n(\text{true})$

Let us perform the fixed point iteration for φ as:

$$0. \quad \varphi^0(\text{true}) \Rightarrow \text{true}$$

$$1. \quad \begin{aligned} \varphi^1(\text{true}) &= (x=0 \wedge f) \vee (x \neq 0 \wedge \text{wfp}[\boxed{x := x-2}](\text{true})) \\ &= (x=0 \wedge f) \vee (x \neq 0 \wedge \text{true}) \\ &= (x=0 \wedge f) \vee (x \neq 0) \end{aligned}$$

We can consider $f: x := 2n+1$, $n \in \text{Integer}$

i.e. $x := \boxed{1, 3, 5, \dots}$

$$\begin{aligned} \varphi^1(\text{true}) &\stackrel{\text{Assume } x := 1 \text{ wlog}}{=} (x := 0 \wedge x \neq 1) \vee (x \neq 0) \\ &= (x \neq 0) \end{aligned}$$

$$\begin{aligned} \varphi^2(\text{true}) &= (x=0 \wedge f) \vee (x \neq 0 \wedge \text{wfp}[\boxed{x := x-2}](x \neq 0)) \\ &= (x=0 \wedge x=1) \vee (x \neq 0 \wedge x-2 \neq 0) \\ &= \text{false} \vee (x(x-2) \neq 0) \\ &= x(x-2) \neq 0. \end{aligned}$$

In this way we'll get $x(x-2) \dots (x-2n) \neq 0$.

Now, we can see that P does not terminate if $x(x-2) \dots (x-2n) \neq 0$, which is true for all $x = \text{odd numbers!}$ i.e. $x := 2n+1$, $n \in \mathbb{I}$.

② Rules for constructing the wlp (weakest liberal pre-condition) gives us that for atomic program lines skip & assignment $x := E$ the rules are exactly the same. For the remaining programs without the existence of loop, definitions differ only in the fact that we use wlp instead of wp on the program lines. So, in case of non-termination which only happens for loops both wp and wlp are structurally same.

So, for a loop-free P \vdash_f , $wp(P, f) = wlp(P, f)$.

Now when we look at while rule we can prove:

$$\text{③ } \left\{ \begin{array}{l} lfp_x. ((\varphi \wedge wlp[P](x)) \vee (\neg \varphi \wedge f)) \\ \star \\ \Sigma gfp_x. ((\varphi \wedge wlp[P](x)) \vee (\neg \varphi \wedge f)) \end{array} \right.$$

which actually will correspond to what we want to prove (by definition)

$$\text{④ } wp(P, f) \leq wlp(P, f)$$

using the hint in the question we know that wlp is monotonic, thus by using Knaster Tarski, we know all the fixed points are forming a complete lattice thus

The order \star holds proving the order ④ we were asked.

