

1. Bayesian Inference

$P(\text{pass} \text{Smart}, \text{Study})$	$\neg\text{smart}$	smart
$\neg\text{study}$	0.2	0.7
study	0.6	0.95

a) $P(\text{Smart}, \text{Study}, \text{Pass}) = P(\text{Smart}) \times P(\text{Study}) \times P(\text{Pass}|\text{Smart}, \text{Study})$

b) JPT Table for Smart, Study, Pass:

	$\neg\text{smart}$		smart	
	$\neg\text{study}$	study	$\neg\text{study}$	study
$\neg\text{pass}$	0.336	0.112	0.054	0.006
pass	0.084	0.168	0.126	0.114

c) $P(\neg\text{study}, \text{pass}) = P(\text{smart}, \neg\text{study}, \text{pass}) + P(\neg\text{smart}, \neg\text{study}, \text{pass})$
 $= 0.126 + 0.084 = 0.21$

$$P(\text{smart}|\neg\text{study}, \text{pass}) = P(\text{smart}, \neg\text{study}, \text{pass}) / P(\neg\text{study}, \text{pass})$$

$$= 0.126 / 0.21 = 0.6$$

$$P(\text{smart}|\neg\text{study}, \text{pass}) = 0.6$$

d) $P(\text{smart}, \neg\text{pass}) = P(\text{smart}, \neg\text{study}, \neg\text{pass}) + P(\text{smart}, \text{study}, \neg\text{pass})$
 $= 0.054 + 0.006 = 0.06$

$$P(\neg\text{study}|\text{smart}, \neg\text{pass}) = P(\text{smart}, \neg\text{study}, \neg\text{pass}) / P(\text{smart}, \neg\text{pass})$$

$$= 0.054 / 0.06 = 0.9$$

$$P(\neg\text{study}|\text{smart}, \neg\text{pass}) = 0.9$$

e) $P(\text{pass}|\text{smart}) = P(\text{smart}, \neg\text{study}, \text{pass}) + P(\text{smart}, \text{study}, \text{pass})$
 $= 0.126 + 0.114 = 0.24$

$$P(\text{pass}|\text{smart}) = 0.24$$

f) $P(\text{pass}|\text{study}) = P(\neg\text{smart}, \text{study}, \text{pass}) + P(\text{smart}, \text{study}, \text{pass})$
 $= 0.168 + 0.114 = 0.282$

$$P(\text{pass}|\text{study}) = 0.282$$

2. Bayesian Networks

a) $P(\text{Cold}, \text{Sneeze}, \text{Allergic}, \text{Scratches}, \text{Cat}) = P(\phi) = \sum_{\omega \in \phi} P(\omega)$

$$P(\text{Cold}, \text{Sneeze}, \text{Allergic}, \text{Scratches}, \text{Cat}) = P(\text{Cold}) \times P(\text{Cat}) \times P(\text{Allergic}|\text{Cat})$$

$$\times P(\text{Sneeze}|\text{Cold}, \text{Allergic}) \times P(\text{Scratches}|\text{Cat})$$

b) $P(\neg\text{cold}, \text{sneeze}, \text{allergic}, \text{scratches}, \text{cat}) = 0.95 * 0.02 * 0.75 * 0.7 * 0.5$
 $= 0.005$

c) $P(\text{cat}|\neg\text{cold}, \text{sneeze}, \text{allergic}, \text{scratches}) = 0.995$

d) $P(\text{cat}|\text{scratches}) = 0.1695$

e) $P(\text{scratches}) = \sum(P(\text{scratches}|\text{cat}) \times P(\text{cat})) + \sum(P(\text{scratches}|\neg\text{cat}) \times P(\neg\text{cat}))$

3. PDDL and Situation Calculus

a) PDDL operator:

```
(:action start-car
  :parameters (?car - car ?key - key ?loc - location)
  :precondition (and
    (at ?car ?loc)
    (has-key ?key)
    (battery-charged ?car)
    (has-gas ?car))
  :effect (and
    (car-running ?car)
    (not (battery-charged ?car))
    (decrease (gas-level ?car) 1))
)
```

b) Situation Calculus:

Action: StartCar(c, s)

Preconditions:

$At(c, loc, s) \wedge HasKey(key, s) \wedge BatteryCharged(c, s) \wedge HasGas(c, s)$

Effects:

$CarRunning(c, do(StartCar(c, s))) \wedge$
 $\neg BatteryCharged(c, do(StartCar(c, s))) \wedge$
 $\neg HasGas(c, do(StartCar(c, s)))$

c) Frame Axiom:

$\forall c, c_2, s (c \neq c_2) \rightarrow (HasGas(c_2, s) \leftrightarrow HasGas(c_2, do(StartCar(c, s))))$