# Machine Intelligence

**Assignment 3** 

(Chapters Slides)

#### Team 1

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- Why may agents be uncertain of their state?
  - 1- Partial Observability 2- Nondeterminism 3- Laziness 4- Ignorance
- Why is probability theory is the best way to summarize uncertainty?
  - 1- Numeric 2- Easily updatable 3- Vectorization 4- basic component of Decision Theory
- Why is full joint probability distribution ineffective at all? And how to solve this problem?

Not scalable with number of RVs of interest (exponential complexity). Solution: the concept of independence between RVs to factor the full joint distribution into separate joint distributions.

- Why is Bayes' Rule important?
  - 1- Evaluates diagnostic based on causal and prior knowledge of cause existence in an online fashion.
  - 2- Combines evidences to evaluate posteriors.

- Bayesian Networks represents independence relationships between random variables.
- Each variable has a conditional distribution that is represented in a table CPT.
- Each CPT entry has conditional probability for at most *k* parents.
- Exact inference methods use enumeration to calculate P(Query | event)
- Variable elimination is applied to speedup exact inference in multiply connected trees.
- Approximate inference using direct sampling, rejection sampling and Monte carlo methods is used
  to eliminate the complexity of exact inference.
- First order logic and Relational Probability models are using to construct well-defined distributions.
- Other methods can be used for probabilistic reasoning such as rule-based systems and fuzzy logic.

- Representation of probabilistic temporal (dynamic) processes:
  - 1) States and observations. 2) Transition and sensor models.
- Inference in probabilistic temporal (dynamic) processes:
  - 1) Filtering. 2) prediction. 3) Smoothing. 4) Most likely explanation.
- Three families of temporal models :
  - 1) Hidden Markov Models (HMMs). 2) Kalman Filters.
  - 3) Dynamic Bayesian Networks (DBNs).
- Data association: keeping track of many objects.

- Reinforcement Learning uses a feedback (reward) from environment to learn and take actions.
- Passive Reinforcement Learning can be done through:
  - 1) Direct Utility Estimation.
  - 2) Adaptive Dynamic Programming.
  - 3) Temporal-Difference (TD) Learning.
- Active Reinforcement Learning:
  - 1) Exploration vs. Exploitation.
  - 2) Q-Learning & SARSA.
- Function Approximation for generalization and dealing with large state spaces.
- Policy Search through policy gradient (differentiable) or empirical gradient (hill climbing).
- Reinforcement Learning for game playing and robot control.