

# Reward is enough

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# Introduction

# What is Intelligence ?

- One of the fundamental question of science
  - Natural Intelligence : how to understand existing intelligence
  - Artificial Intelligence : How to implement & make new Intelligence
- Expressions of intelligence in animal and human behaviour are so bountiful and so varied
  - social intelligence, language, perception, motor control ..
- What could drive agents (natural or artificial) to behave intelligently in such a diverse variety of ways?
  - how can it be that same animal both has knowledge, social intelligence, perception .. ?
  - how can we understand these rich diversity of Intelligence ?

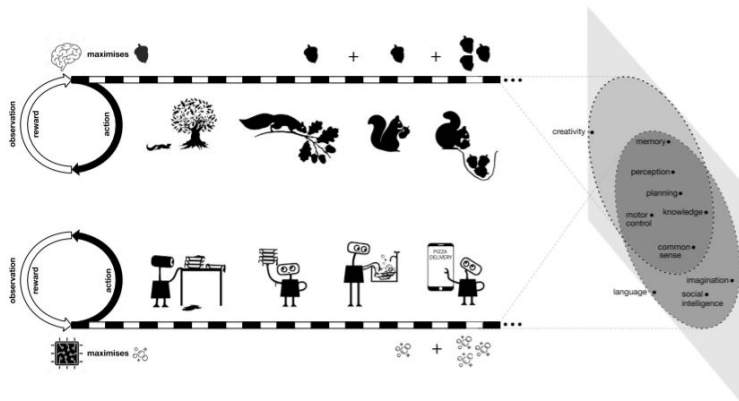
# What is Intelligence ?

- One possible Answer : each ability arises from the pursuit of specialized goal (specialized problem formulation)
- Traditional approach to AI : break up AI into each field of Intelligence
  - Perception is driven by object recognition
  - Language is driven by tagging, parsing, sentiment analysis, next word prediction(GPT3)
  - Social intelligence is driven by the pursuit of a Nash equilibrium
  - ..

# Goal

- Intelligence may pursue a variety of different goals
  - An animal may act to minimise hunger
  - A go playing agent may act to maximise win
- Reward provides a flexible representation of goals
- whatever the goal it is, Reward itself is enough to understand intelligence
  - why do we need separate goal and framework to understand them?

# Example



- the pursuit of one goal may generate complex behaviour that exhibits multiple abilities associated with intelligence.

## Many Attributes, Common Goal

- A complex environment demands many attributes of intelligence
  - ex) perception, languages, social intelligence may be necessary to survive
- Reward maximization may then be viewed as a common goal
  - Perception, language, social intelligence emerge in the pursuit of reward maximization
- Reward maximization may be sufficient to understand intelligence
  - Explaining the diverse expressions intelligence found in natural intelligence



# Reward is Enough Hypothesis

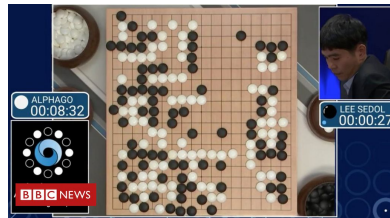
# Reward is Enough Hypothesis

- Note : the reward hypothesis speculates that all goals may be represented by rewards
  - This should not be confused with our reward-is-enough hypothesis
- Sophisticated abilities may arise from the maximisation of simple rewards in complex environments

## Hypothesis (Reward-is-Enough):

“ All attributes of intelligence can be understood as subserving the maximisation of reward by agent acting in its environment ”

## Example : Compute go



- Prior work focoused on seperate goals
  - Shape (pattern recognition)
  - Tactics (local search)
  - Endgames (combinatorial game theory)
- AlphaGo focoused on a common goal of maxmising game won
  - Led to deeper understanding of shape, tactics, endgames ..
  - Produced a broder set of Intelligence
    - Territory and Influence, thickness, attack-defence
- Integrated into a unified Intelligence

## Why Reinforcement learning? (1)

- Reinforcement learning defines the problem of reward maximisation
- RL address the practical problem faced by natural and artificial intelligence
  - Not a theoretical abstraction of the problem
- Agent is bounded by practical constraints
  - Agent has limited capacity determined by its machinery
    - Limited neurons in brain
    - Limited memory in a computer
  - Agent runs in real-time
    - synapses take time to fire
    - CPU takes time to process

∴ we need to solve reward maximisation problem subject to these practical constraints

## Why Reinforcement learning? (2)

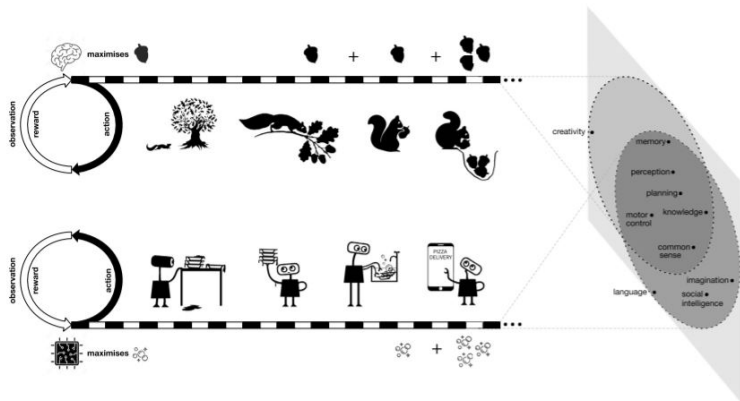
The definition of environment is broad and encompasses many problem dimensions.

Dimension	Alternative A	Alternative B	Notes
Observations	Discrete	Continuous	Time-step may be infinitesimal
Actions	Discrete	Continuous	
Time	Discrete	Continuous	
Dynamics	Deterministic	Stochastic	
Observability	Full	Partial	Other agents are part of environment, from perspective of single agent (see Section 3.3)
Agency	Single agent	Multi-agent	
Uncertainty	Certain	Uncertain	
Termination	Continuing	Episodic	Environment may terminate and reset to an initial state
Stationarity	Stationary	Non-stationary	Environment depends upon history and hence also upon time
Synchronicity	Asynchronous	Synchronous	Observation may remain unchanged until action is executed
Reality	Simulated	Real-world	May include humans that interact with agent

- Environment definition is incredibly general
- it allows vast range of different problem to be expressed

# Reward is Enough for ..(somtehing)

# Reward is Enough for ..(something)



- Reward is enough for knowledge and learning
- Reward is enough for perception
- Reward is enough for language
- ..

## Reward is Enough for Knowledge and Learning

- Some environment demand prior knowledge
  - A newborn gazell may have to run away from a lion
  - Prior knowledge limited by agent's capacity, and difficulty to construct
- Some environmnet demand learned knowledge
  - The future is uncertain (unkown/ stochasticity / complexity)
  - Total space of potential knowledge » capacity of agent
    - lions in savannah / polar bears
  - knowledge must adapt to circumstance
- in complex environment, Learned knowledge will become increasingly important
- ∴ Reward is Enough for Knowledge and Learning



# Reward is Enough for Perception

- Rich real world environment may demand various perceptual skills
  - Image segmentation ( avoiding falling off a cliff )
  - Object recognition ( to identify good and bad foods )
  - Face recognition ( to differentiate friend from foe )
  - Speech recognition ( to understand verbal warning )
- Alternatively suggest that perception may instead be understood as subserving the maximisation of reward.
- RL Problem include richer variety of perceptual task
  - There is no labelled data
  - Action and observation are intertwined ( haptic perception identifying content of pocket )
  - Utility is policy dependent ( misclassifying a crocodile when swimming vs walking)
  - Distribution of data is context dependent
    - classifying ice and polar bears vs savanah and lions

## Reward is Enough for Language

- Advances have come from viewing languages as solution to a common goal
  - Predicting the next word in a large corpus of data (GPT3)
- Language modelling may not produce broader linguistic Intelligence
  - Language may be intertwined with other modalities of action and observation
    - two agents discussing how to carry an awkward object
  - Language is situational and consequential and tailored to achieve a purpose
    - Salesperson maximize sales , politician maximize votes
- Richer language may emerge from a common goal of reward maximization
  - An agent must learn to comprehend 'danger' warning to survive ..
- ∴ The ability of language in its full richness, including all of these broader abilities, arises from the pursuit of reward.

# Discussion

## What Else Could be Enough

- Unsupervised Learning
  - Provides no goals for action-selection
  - it can't be enough itself for intelligent behavior
- Supervised Learning
  - it can be enough to achieve the goal that the teacher want to
  - but Can only address teacher's environment and goal
- Offline learning from Large Data-set
  - real problems in complex environment far larger than any data-set
  - Data-set of chimp playing Lego not enough to build a bridge
- ..

" All attributes of intelligence can be understood as subserving the maximisation of reward by agent acting in its environment "

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