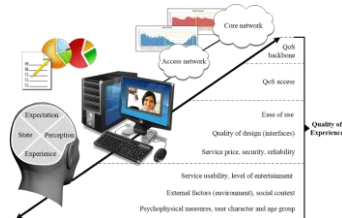


FLEXIBLE USER MODEL FOR HUMAN'S COGNITIVE JUDGMENT IN VIDEO STREAMING APPLICATIONS

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Background - Quality of Experience (QoE)

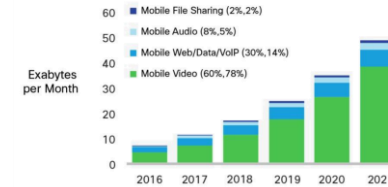
- From QoS to QoE
 - QoE is a measure of the overall level of user satisfaction for application
 - Improvement of QoS is not necessary improvement of QoE for a user
 - QoE improvement for user by user is more important aspect



[2] <http://www.fpz.unizg.hr/qoe4vr/index.php/2017/06/26/what-is-quality-of-experience-qoe/>

Background - mobile data traffic explosion

- Rapid increase in the volume of traffic by mobile device^[1]
 - From 2016 to 2021, the volume will be sevenfold
 - Video data traffic will account for 75% of the total traffic



⇒ Guarantee of Quality of Service (QoS) in mobile networks becomes more difficult

[1] Cisco, "Cisco Visual Networking Index: Global Mobile Data Traffic Forecast Update, 2016–2021."

Research Goal

- Problem
 - QoE degradation by QoS fluctuation and user's operation
 - e.g. A user watching a video sometimes underestimates the throughput and the delay time and selects higher bitrate video, which degrades user's own QoE
 - Misperception from cognitive limitations of human causes wrong decision
- Goal
 - QoE improvement by detecting and preventing user's wrong operation caused by recognition error of QoS

As a first step, we propose a human's cognition model that reflects misperception characteristics of human to analyze its behavior in video streaming applications

Approach

- 3 steps for creating human's cognition model in video streaming applications
 1. We clarify the reason of why misperception in human brain
 2. We propose a user agent model where the agent observes information about network performance, estimates the probability that observed performance is obtained, and makes some decisions on the operation for a video streaming application
 3. We reveal that the model proposed in step 2. has the misperception features of step 1.

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Features in human perception

- Two factor of human perception
 - Probability estimation based on the Bayesian estimation
 - **Decision making from experiences**
 - Decision making from small samples^[3]
 - Unwilling to take much time to search for information causes limited short-term memory for decision making
 - Decision making with noisy retrieval from brain memory^[4]
 - Human sometimes makes a decision using relevant and irrelevant memory
 - Erroneous decision is made by over-estimation or under-estimation of probability caused by the above 2 reasons

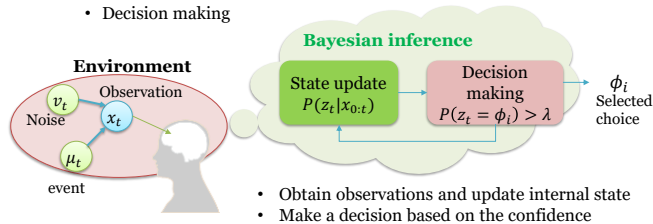
[3] Ralph Hertwig, Timothy J. Pleskac, "Decisions from experience: Why small samples?" *Cognition*, vol. 115, no. 2, pp. 225-237, 2010.

[4] D. Marchioni, S. Di Guida, and I. Erev, "Noisy retrieval models of over- and under-sensitivity to rare events," *Decision*, vol. 2, no. 2, pp. 82-106, 2015.

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Human's cognition model

- Bayesian Attractor Model (BAM)^[5]
 - represents human-brain's perceptual decision making
 - Event sensory
 - Probability estimation for the event
 - Decision making



x_t : Observation
 z_t : State
 ϕ_i : i^{th} choice
 λ : Confidence threshold

- Obtain observations and update internal state
- Make a decision based on the confidence

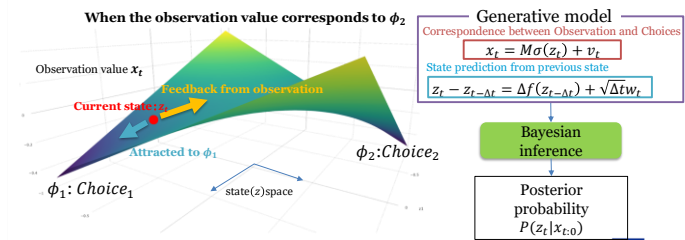
[5] S. Bitzer, et al., "A Bayesian attractor model for perceptual decision making," *PLoS Computational Biology*, vol. 11, no. 8, Aug. 2015.

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State update in BAM

- Update the state every time BAM obtains observation value
 - To approach decision-making state corresponding to observation
 - To be attracted to the choice close to the current state

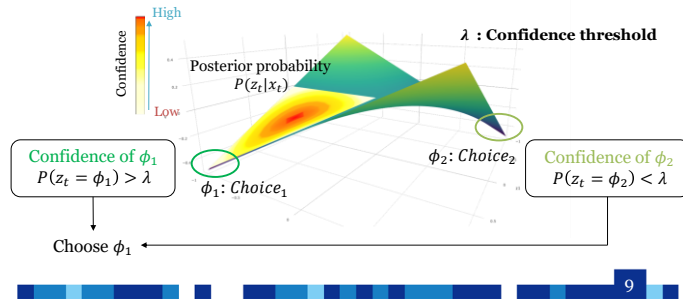
M : Observation matrix
 σ : Sigmoid function
 f : Hopfield dynamics
 v_t, w_t : Noise



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Decision making in BAM

- Calculate posterior probability as a confidence of a choice
- Choose a choice with higher confidence than threshold λ



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Simulation for validation

- Run simulation of BAM to see if a feature appearing in people who make a decision based on small observation information can be captured by BAM or not
 - The feature is “The smaller the observation information is, the lower the accuracy rate becomes”[3]
 - The accuracy rate corresponds with the height of confidence in BAM
 - The number of observation information corresponds with the time taken for decision-making
- Consider the correspondence between the two tradeoffs
 - The smaller the observation information is, the lower the accuracy rate becomes
 - The shorter the time taken for decision-making is, the lower confidence threshold becomes

[3] Ralph Hertwig, Timothy J. Plekac, “Decisions from experience: Why small samples?” Cognition, vol. 115, no. 2, pp. 225-237, 2010.

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Simulation scenario

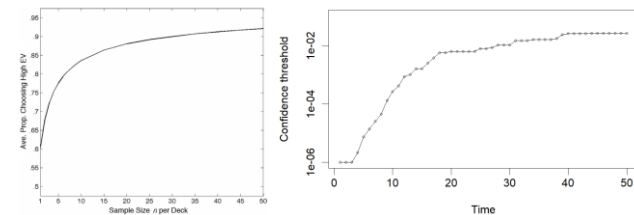
- The feature “The smaller the observation information is, the lower the accuracy rate becomes” [3]
 - A problem of choosing higher payoff distributions (decks) from two decks.
 - Each deck consisting of two outcomes of the type a probability p to win amount x ; otherwise win amount y
 - (x, p, y) randomly sampled from certain range
- BAM
 1. Prepare an environment where there are three choices.
 2. At first, BAM receives observation value corresponding to ϕ_1 , and the state is shifted to a state where ϕ_1 is adopted.
 3. After this, BAM receives Observation value corresponding to another choice B. The confidence of ϕ_2 changes.
 4. Then, the time taken for making decision ϕ_2 is measured for various confidence threshold

[3] Ralph Hertwig, Timothy J. Plekac, “Decisions from experience: Why small samples?”

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Result

- Left figure [3]: the rate of correct making-decision for the number of observation in the feature “decision making from small samples”
- Right figure: the time taken for decision-making for the height of confidence threshold in BAM



[3] Ralph Hertwig, Timothy J. Plekac, “Decisions from experience: Why small samples?”

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Conclusion



- We propose a human's cognition model
 - We show that our proposed model show the typical characteristics of human's misperception
 - By adjusting confidence threshold of the model, different behaviors of individual user's can be modelled
- Future work
 - Consider expressing the feature of Decision making with noisy retrieval from brain memory^[4] with BAM
 - Capture the features of human perception by BAM with fitted parameter and prevent human error of misrecognition