

Effect-Driven Selection of Web of Things Services in Cyber-Physical Systems Using Reinforcement Learning

Baek, KyeongDeok and Ko, InYoung

School of Computing,

Korea Advanced Institute of Science and Technology

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Introduction

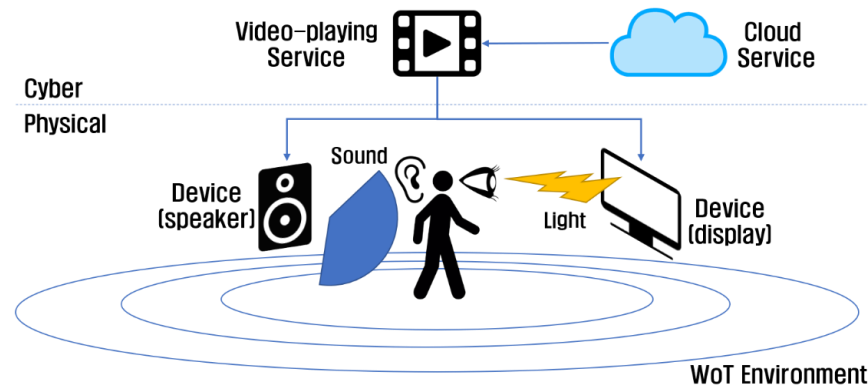
CPS-based Web of Things Environments

■ Web of Things service

- “The **Web of Things (WoT)** is software architectural styles and programming patterns that allow real-world objects to be part of the World Wide Web.” [1]
- Providing services to the user by utilizing connected things

■ Cyber-Physical System (CPS)

- “A **cyber-physical system (CPS)** is a mechanism that is controlled or monitored by computer-based algorithms, tightly integrated with the Internet and its users.” [2]



[1] https://en.wikipedia.org/wiki/Web_of_Things [2] https://en.wikipedia.org/wiki/Cyber-physical_system

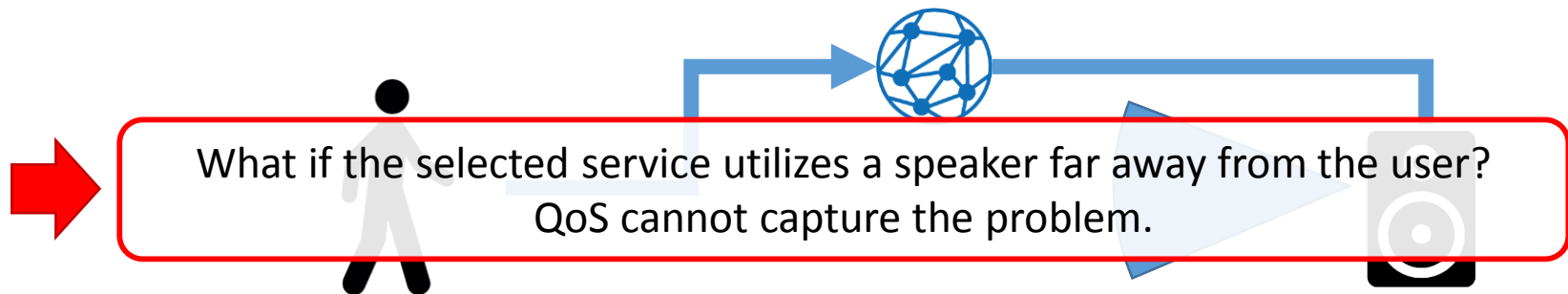
Important Characteristic of WoT Services and Problem

■ Physical effect delivery

- Physical effects (ex. light, sound) generated by things should be delivered to the user and being perceived by the user properly
- User satisfaction of service is affected by quality of effect delivery

■ Web service selection problem

- Selecting best service among functionally equivalent candidates
- In terms of network-level Quality of Service (QoS)
 - Availability, transmission delay, packet loss...

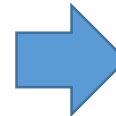


Suggesting Solution

■ Effect-driven selection of WoT services

- Selecting optimal services in terms of quality of physical effects
- The user will evaluate the quality, and a selection algorithm should be able to predict the evaluation

How to measure quality of physical effects in the user's perspective?

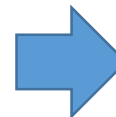


Service effectiveness model

■ Dynamic selection of WoT services

- Dynamically selecting and replacing services to maintain high effectiveness continuously

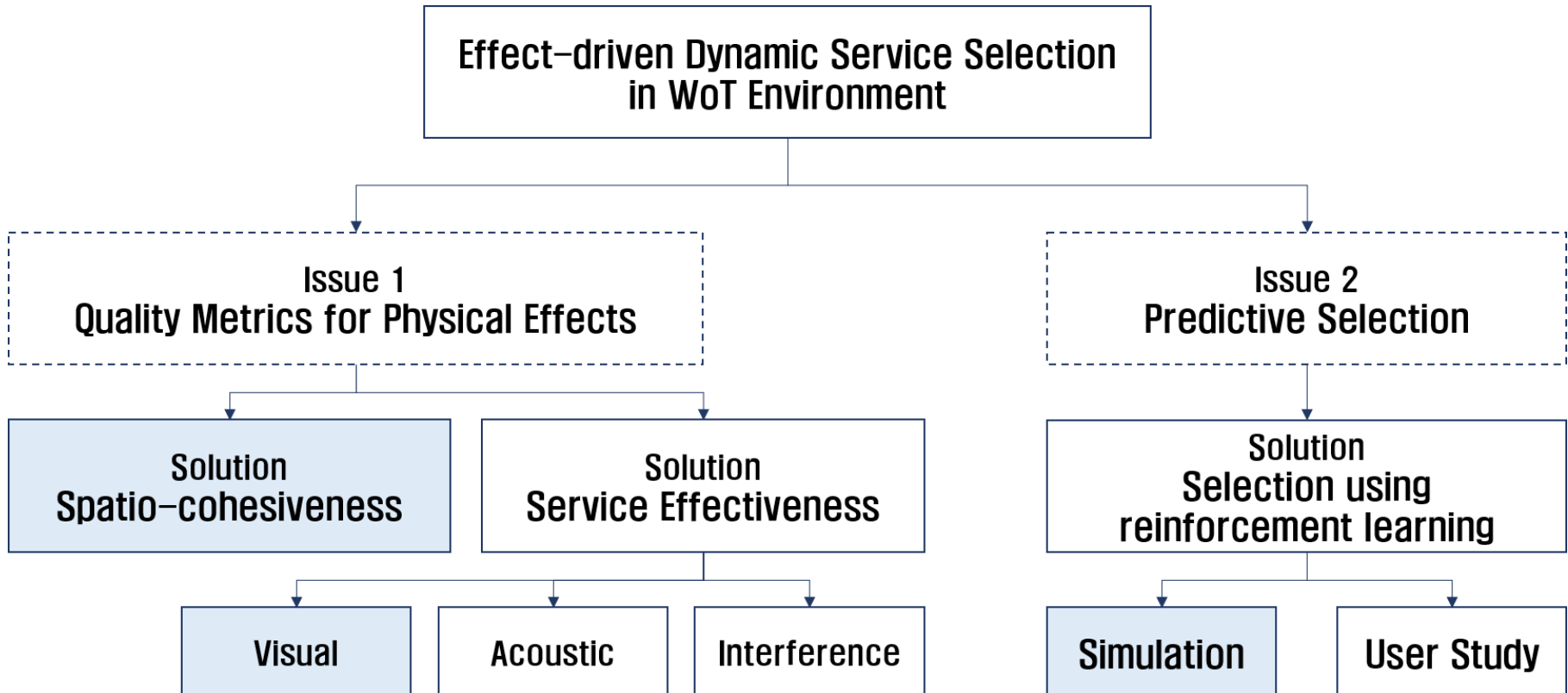
How to select services in a predictive manner to deal with mobile users/devices?



Dynamic service handover

Research Issues

Research Roadmap



Research Issues (Service Effectiveness)

■ Computation model of service effectiveness

- Enable quantitative predicting the service effectiveness
 - Selection is done automatically according to the prediction
- Different service/effect type may have different model
 - Light, sound, heat, ...

■ Service interference model

- Same or different type of physical effects may cause interferences
 - Constructive (synergy)
 - Destructive (disturbance)
- Develop interference-aware effectiveness model for multi-services

Research Issues (Dynamic Service Handover)

■ Dynamic service handover algorithm [1]

- Existing approaches do not consider dynamically changing environments
 - Users and services in WoT environments are highly dynamic
- Replacing service instance within another one to maintain high quality throughout the service provision
- Extension of network-level handover concept to service-level

■ Predictive selection algorithm

- Predict future conditions of candidate services and select the most promising one
 - Reducing the number of service handover

[1] Baek, Kyeong-Deok, and In-Young Ko. "Spatially cohesive service discovery and dynamic service handover for distributed IoT environments." *International Conference on Web Engineering*. Springer, Cham, 2017.

Previous Works

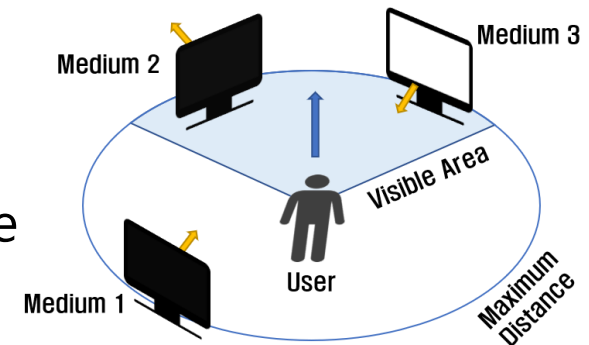
Previous Works (Service Effectiveness)

■ Spatio-cohesiveness [1, 2]

- Measure how the selected devices are spatially cohesive
 - ✓ Assume highly cohesive devices are likely to provide high quality of physical effects, because of locality
- [Limitation] Short distance cannot guarantee effectiveness

■ Visual service effectiveness [3]

- Rule-based model that measure whether a user can read texts on a display
 - ✓ Distance, Field of View (FoV), Facing rule



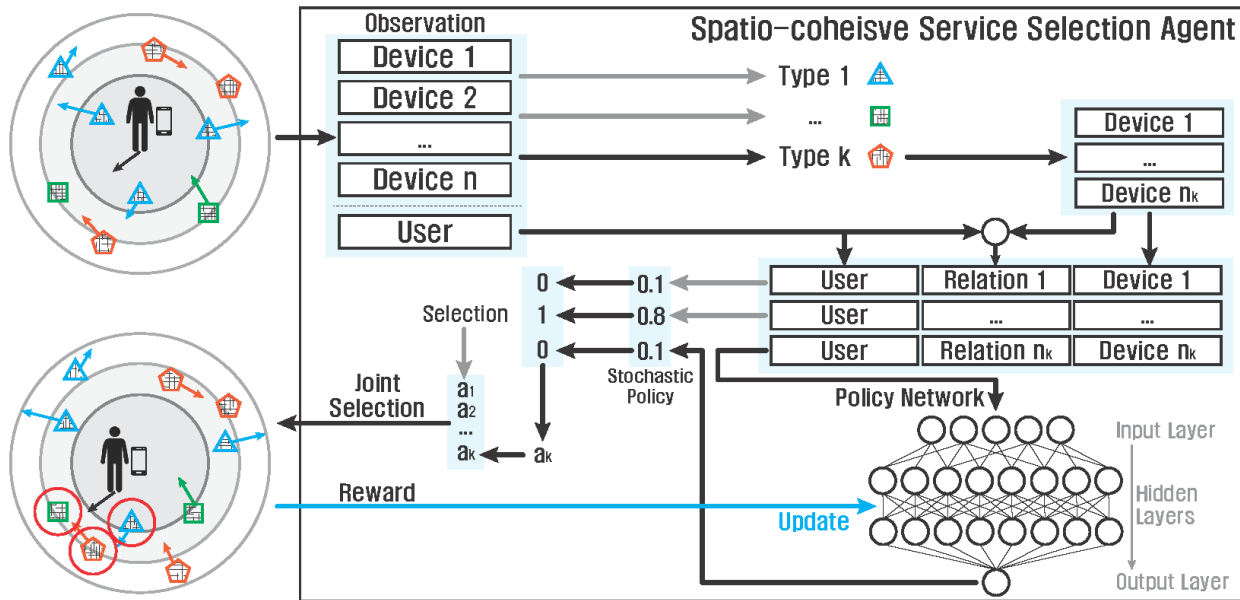
[1] Baek, Kyeong-Deok, and In-Young Ko. "Spatially cohesive service discovery and dynamic service handover for distributed IoT environments." International Conference on Web Engineering. Springer, Cham, 2017.

[2] Baek, KyeongDeok, and In-Young Ko. "Spatio-Cohesive Service Selection Using Machine Learning in Dynamic IoT Environments." International Conference on Web Engineering. Springer, Cham, 2018.

[3] Kyeongdeok Baek, and In-Young Ko, "Effect-driven Dynamic Selection of Physical Media for Visual IoT Services using Reinforcement Learning", To appear in Proceedings of the 24th IEEE International Conference on Web Services (ICWS 2019), July 8 – 13, 2019, Milan, Italy

Previous Works (Dynamic Service Handover)

- **RL-based dynamic service selection (handover) algorithm [1, 2]**
 - Perform predictive and dynamic selection on services
 - Learn from experiences to select most promising service
 - Tried actor-critic algorithm [1] and DQN algorithm [2]

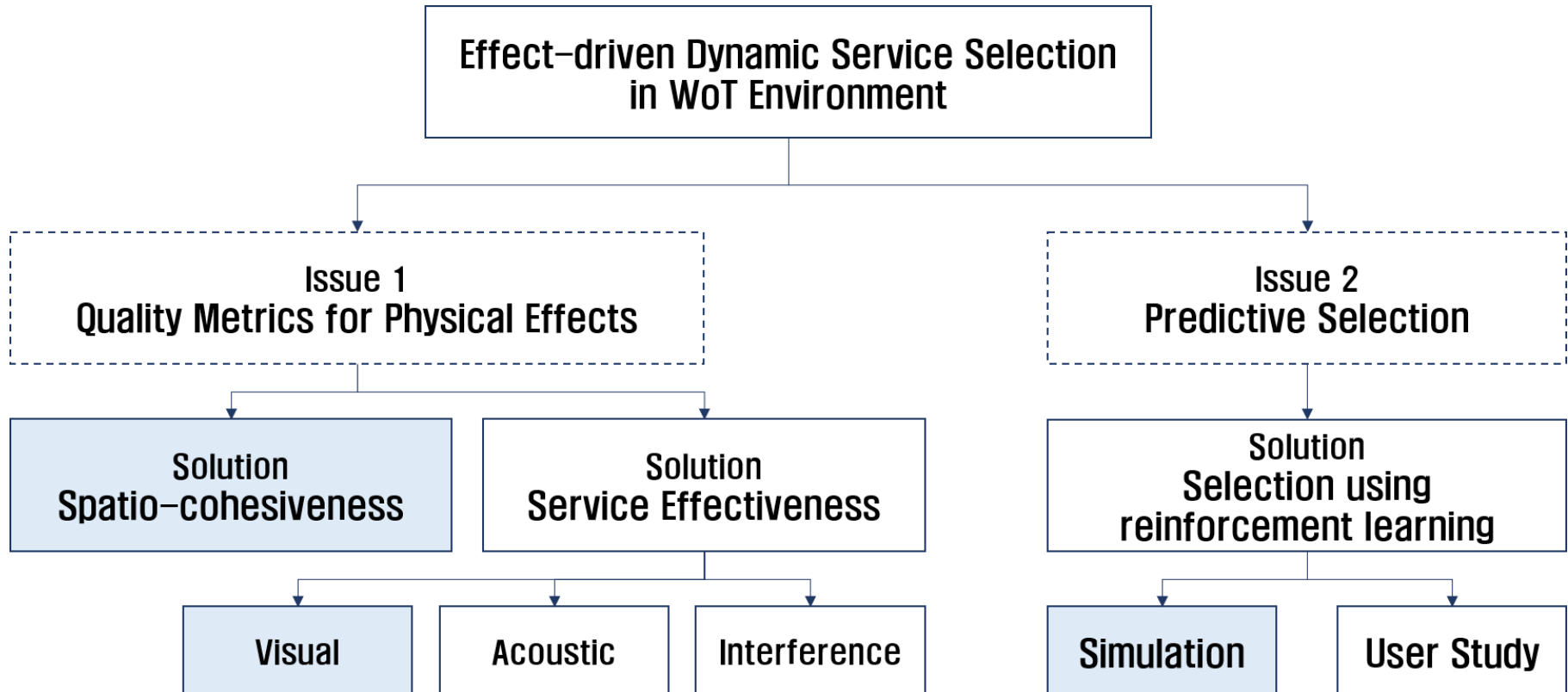


[1] Baek, Kyeong-Deok, and In-Young Ko. "Spatially cohesive service discovery and dynamic service handover for distributed IoT environments." International Conference on Web Engineering. Springer, Cham, 2017.

[2] Kyeongdeok Baek, and In-Young Ko, "Effect-driven Dynamic Selection of Physical Media for Visual IoT Services using Reinforcement Learning", To appear in Proceedings of the 24th IEEE International Conference on Web Services (ICWS 2019), July 8 – 13, 2019, Milan, Italy

Research Plan

Research Roadmap (Remind)



Research Plan (Service Effectiveness)

1. Type-specific service effectiveness model

- Develop service effectiveness model for each type
 - Currently, only visual model is partially developed [1]
- Action items
 - 1) Acoustic service effectiveness model
 - 2) Improve visual service effectiveness model

2. Service interference model

- Analyze service interference among the services that generate similar or different types of physical effects
- Action items
 - 1) Develop interference prediction model

[1] Kyeongdeok Baek, and In-Young Ko, "Effect-driven Dynamic Selection of Physical Media for Visual IoT Services using Reinforcement Learning", To appear in Proceedings of the 24th IEEE International Conference on Web Services (ICWS 2019), July 8 – 13, 2019, Milan, Italy

Research Plan (Dynamic Service Handover)

1. Virtual Reality-powered user study

- Perform user study for
 - Evaluating practicality of our approach
 - Collecting realistic data from real users
- VR-powered simulation for cost-efficiency
 - ICWE 2019 Demo [1]



2. Learn from human preferences

- Rather than using reward signals, guided by human preferences [2]
- Applying latest RL techniques

[1] Baek, KyeongDeok, HyeongCheol Moon, and In-Young Ko. "VR-Powered Scenario-Based Testing for Visual and Acoustic Web of Things Services." International Conference on Web Engineering. Springer, Cham, 2019.

[2] Christiano, P.F., Leike, J., Brown, T., Martic, M., Legg, S., Amodei, D.: Deep reinforcement learning from human preferences. In: Advances in Neural Information Processing Systems, pp. 4299–4307 (2017)

References

- [1] **Baek, Kyeong-Deok**, and In-Young Ko. "Spatially cohesive service discovery and dynamic service handover for distributed IoT environments." International Conference on Web Engineering. Springer, Cham, 2017.
- [2] **Baek, KyeongDeok**, and In-Young Ko. "Spatio-Cohesive Service Selection Using Machine Learning in Dynamic IoT Environments." International Conference on Web Engineering. Springer, Cham, 2018.
- [3] Christiano, P.F., Leike, J., Brown, T., Martic, M., Legg, S., Amodei, D.: Deep reinforcement learning from human preferences. In: Advances in Neural Information Processing Systems, pp. 4299–4307 (2017)
- [4] **Kyeongdeok Baek**, and In-Young Ko, "Effect-driven Dynamic Selection of Physical Media for Visual IoT Services using Reinforcement Learning", To appear in Proceedings of the 24th IEEE International Conference on Web Services (ICWS 2019), July 8 – 13, 2019, Milan, Italy
- [5] **Baek, KyeongDeok**, HyeongCheol Moon, and In-Young Ko. "VR-Powered Scenario-Based Testing for Visual and Acoustic Web of Things Services." International Conference on Web Engineering. Springer, Cham, 2019.
- [6] **Baek, KyeongDeok**, and In-Young Ko. "Effect-Driven Selection of Web of Things Services in Cyber-Physical Systems Using Reinforcement Learning." International Conference on Web Engineering. Springer, Cham, 2019.

Thank you

- Q&A
- Comments