

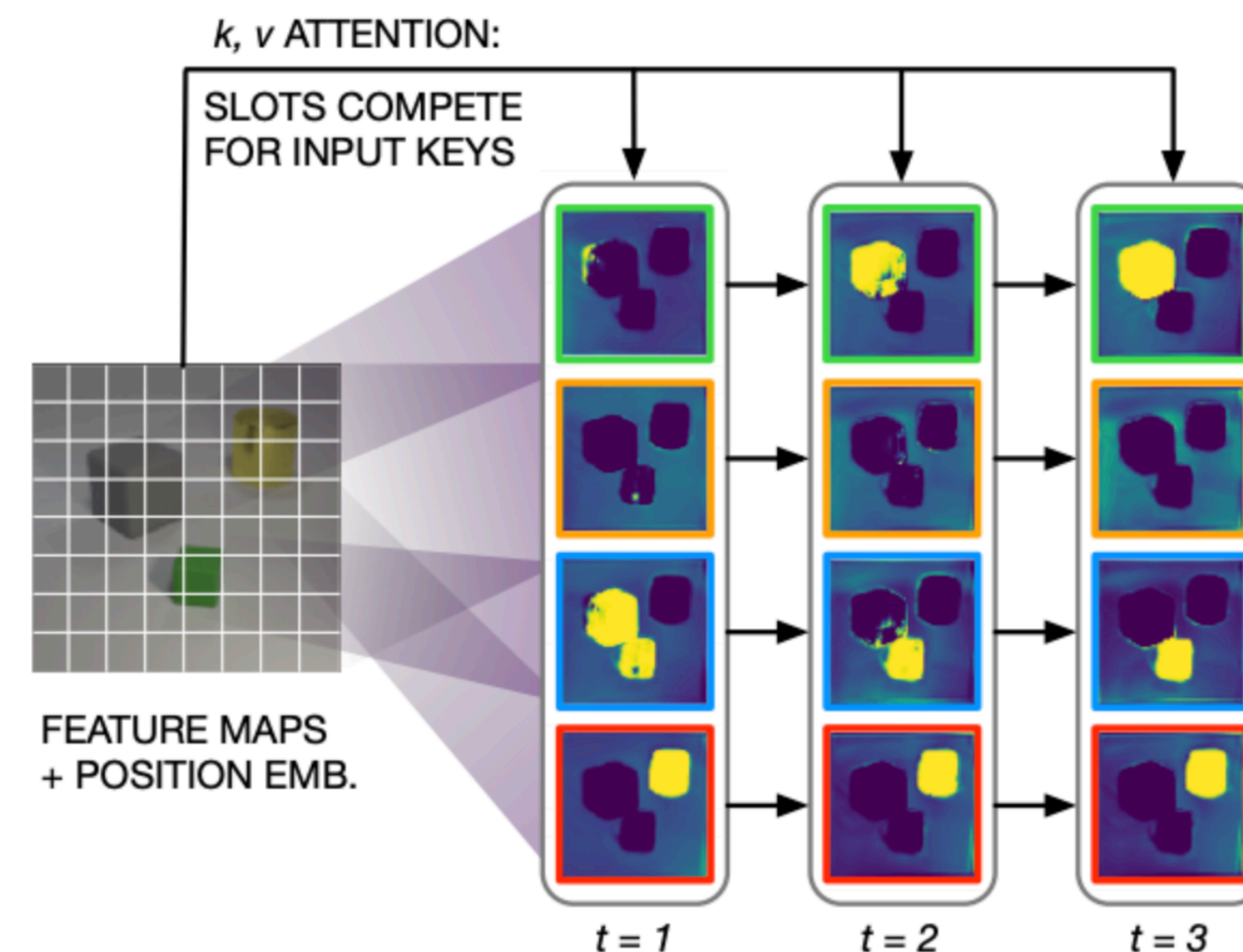
Bootstrapping Top-down Information for Self-modulating Slot Attention

Dongwon Kim, Seoyeon Kim, Suha Kwak
POSTECH

Task: Object-centric learning

What is object-centric learning?

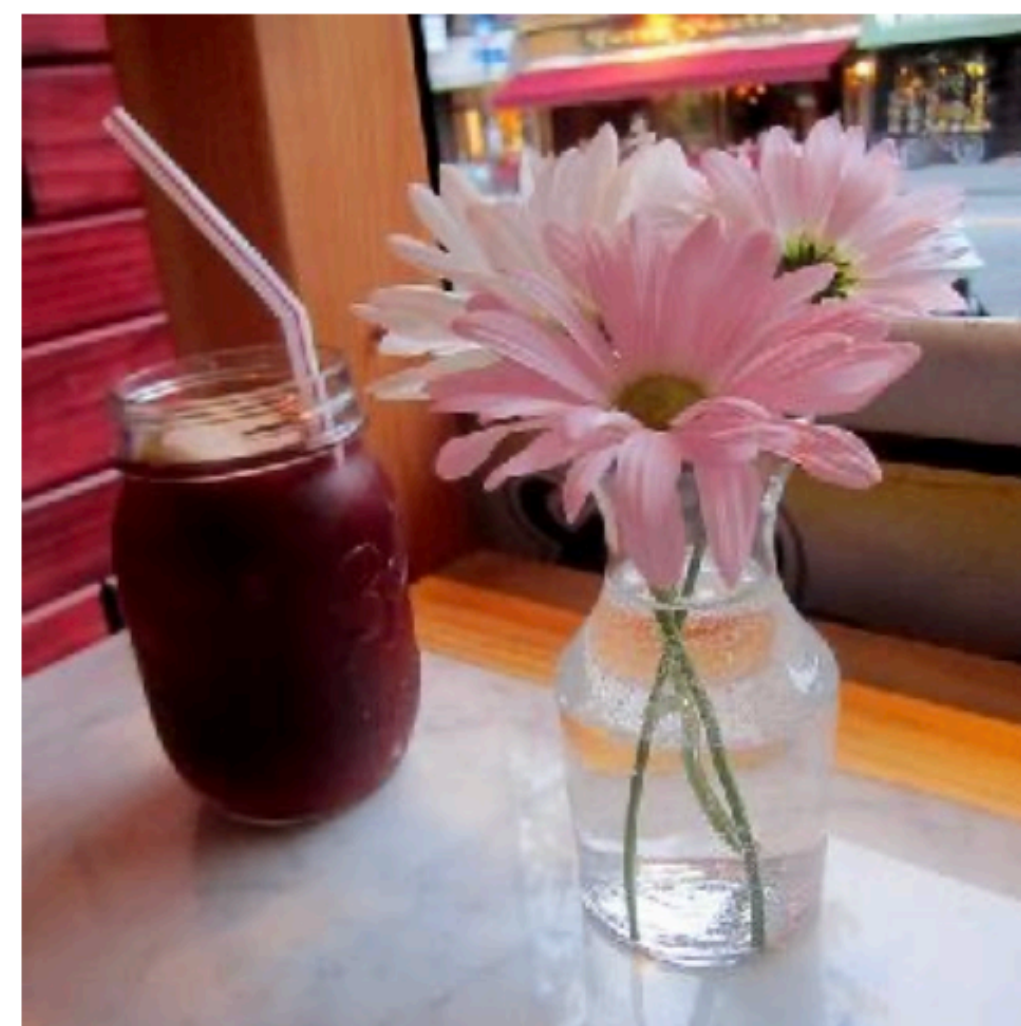
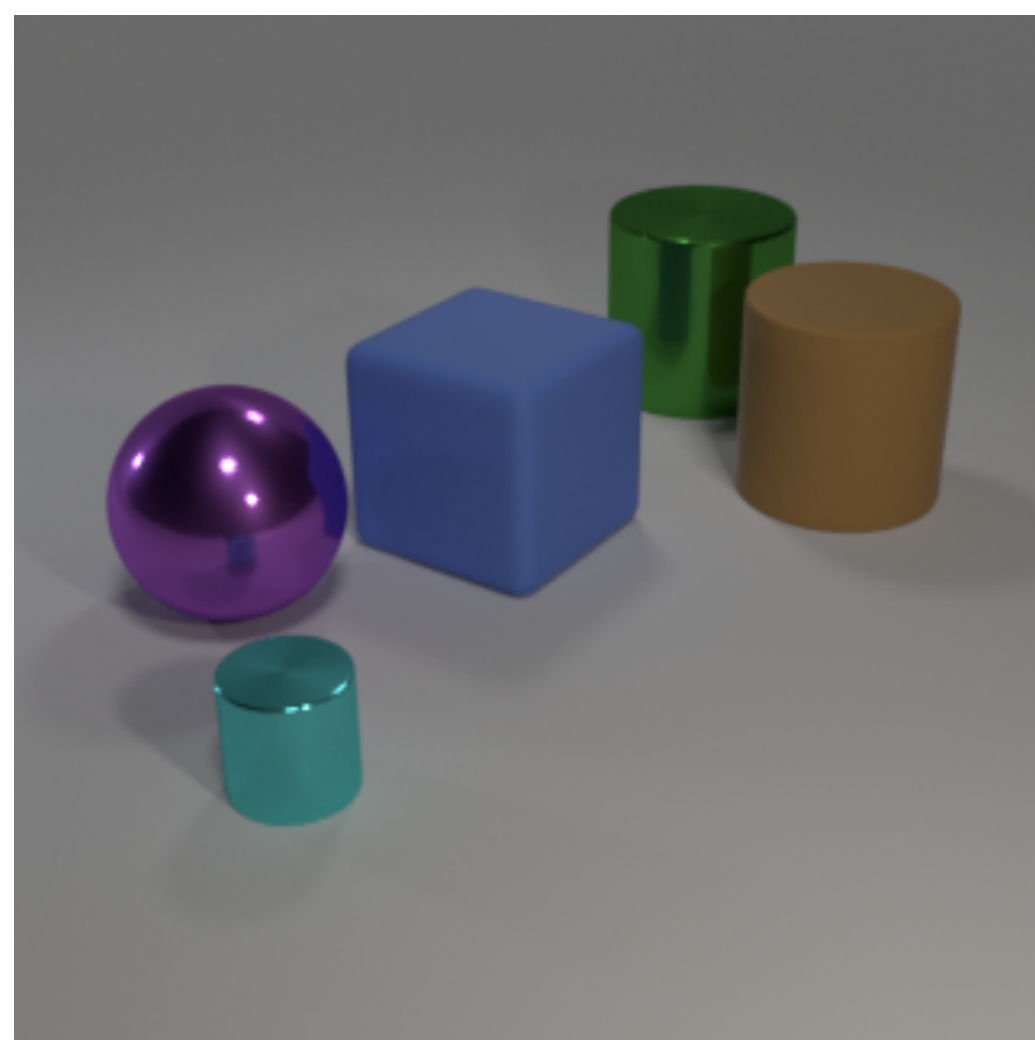
- Task of learning representations of individual objects from visual scenes without manual labels
- Slot-attention: compute per-object representations (slot) through clustering-like attention



Motivation: Incorporating top-down approaches

Conventional OCLs are bottom-up methods

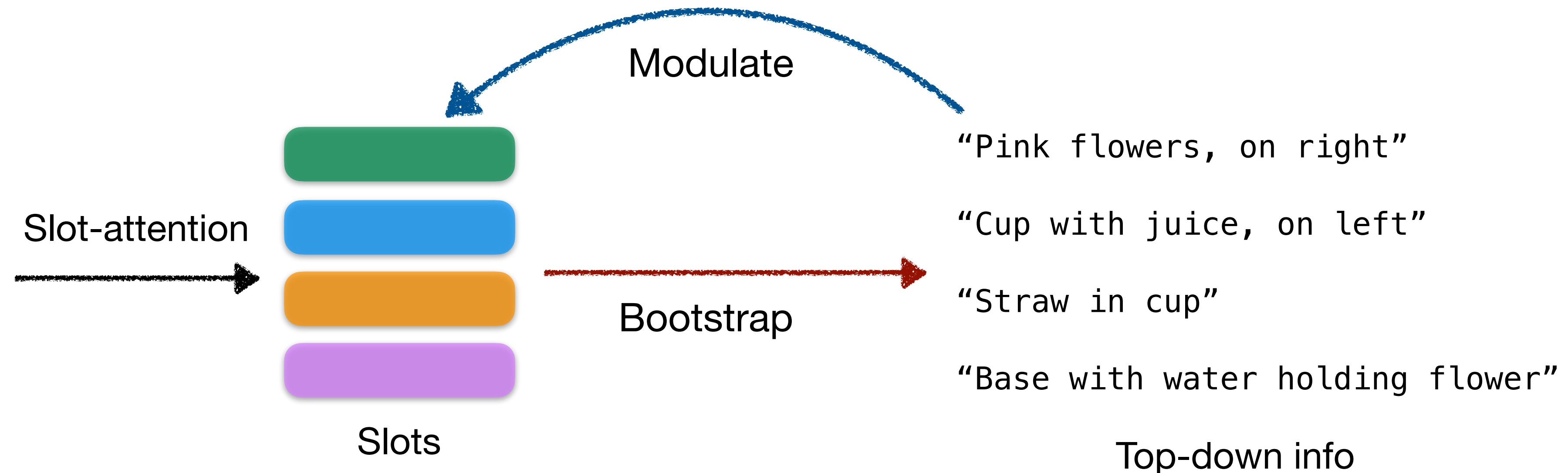
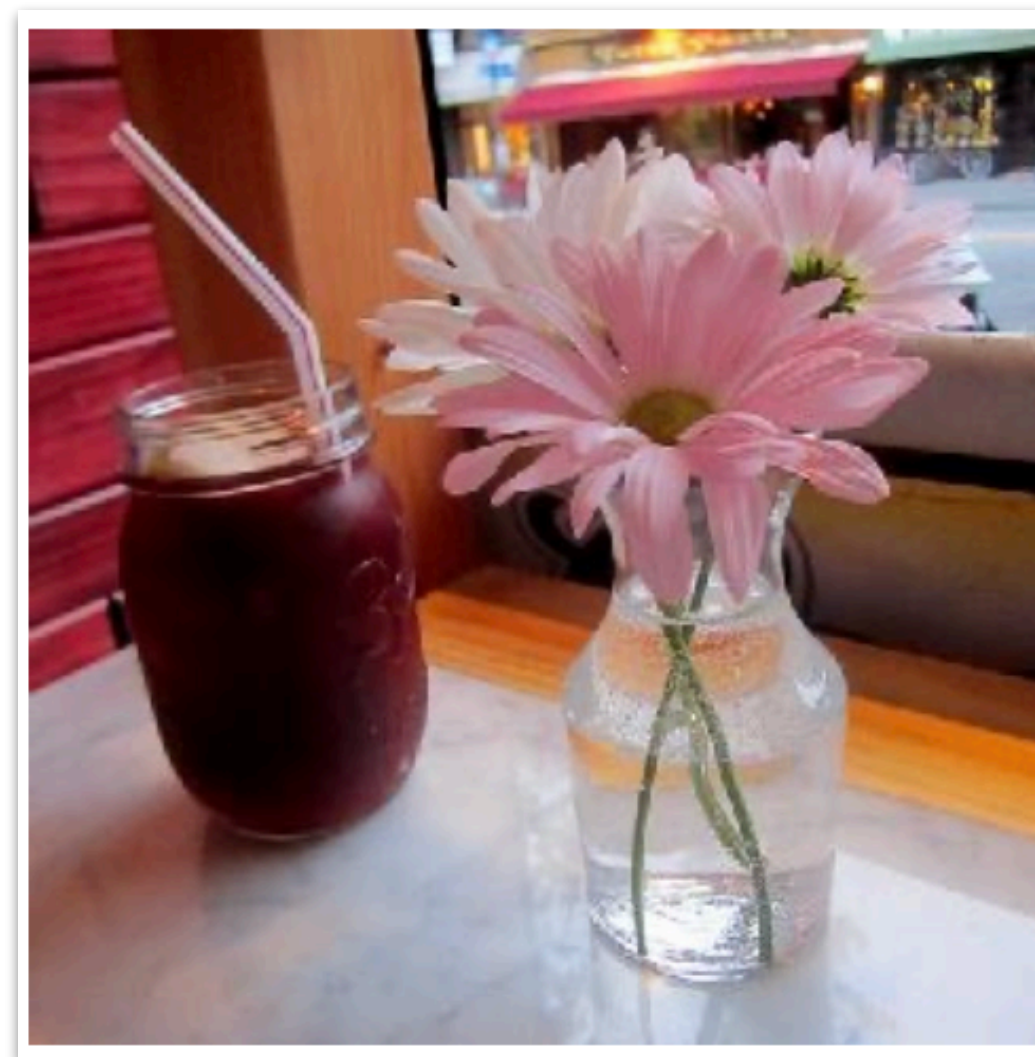
- Slot attention can be considered as a **bottom-up** method; it relies on aggregating features without high-level semantic information.
- It assumes that features within an object are homogeneous & can be clustered.
 - But this does not always hold for the real-world images



Motivation: Incorporating top-down approaches

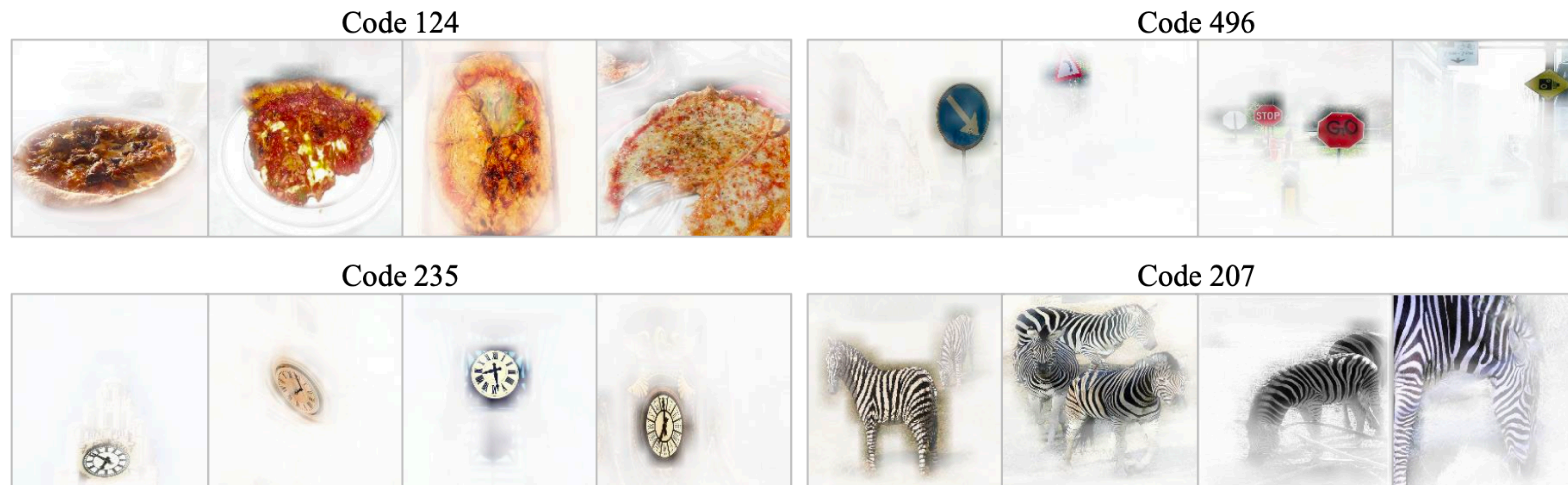
Can we *bootstrap* top-down info, and then use it to *modulate* model?

- Incorporating top-down information can fix the issue
 - Problem will be much easier if we can provide model about the semantic information or rough location of each object



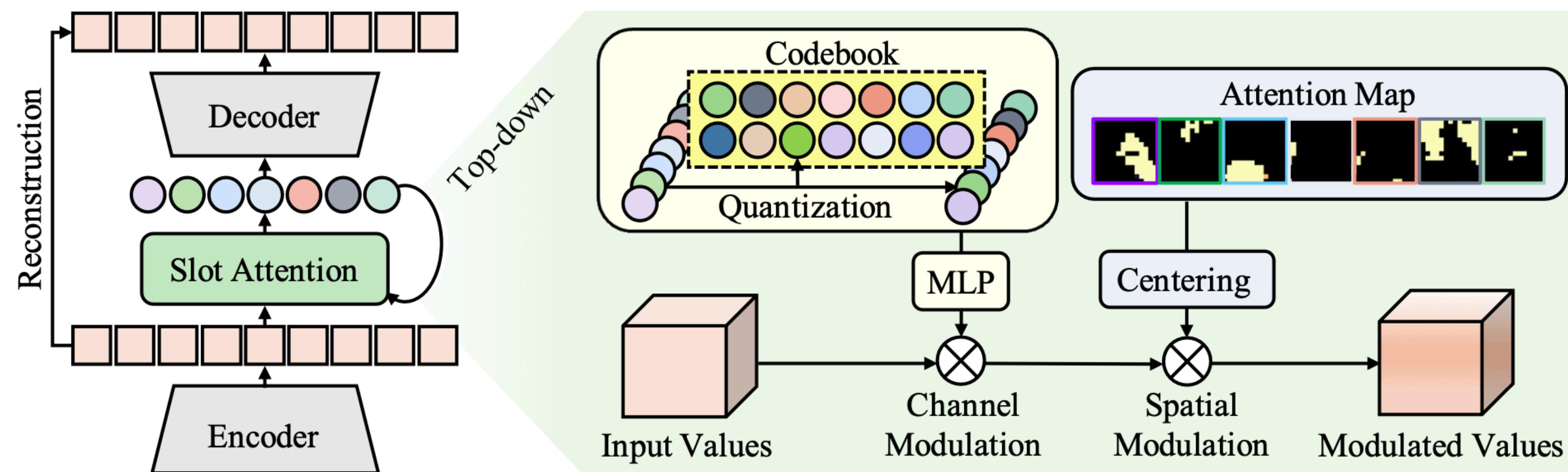
Bootstrapping top-down information

- During training, the codebook of slots is learned together using vector quantization.
- Codebook learns to store distinct semantic patterns, where each code can act as automatically discovered top-down semantic information.



Self-modulating slot attention

- Using bootstrapped top-down information, input value of the slot attention is modulated.
- The modulation guides the update of the slots, by prioritizing visual features relevant to top-down information



Results

- Our proposed top-down pathway largely improved the quality of object-centric representation in the real-world images.



Thank you!

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