

Lecture 7: Graph Neural Networks

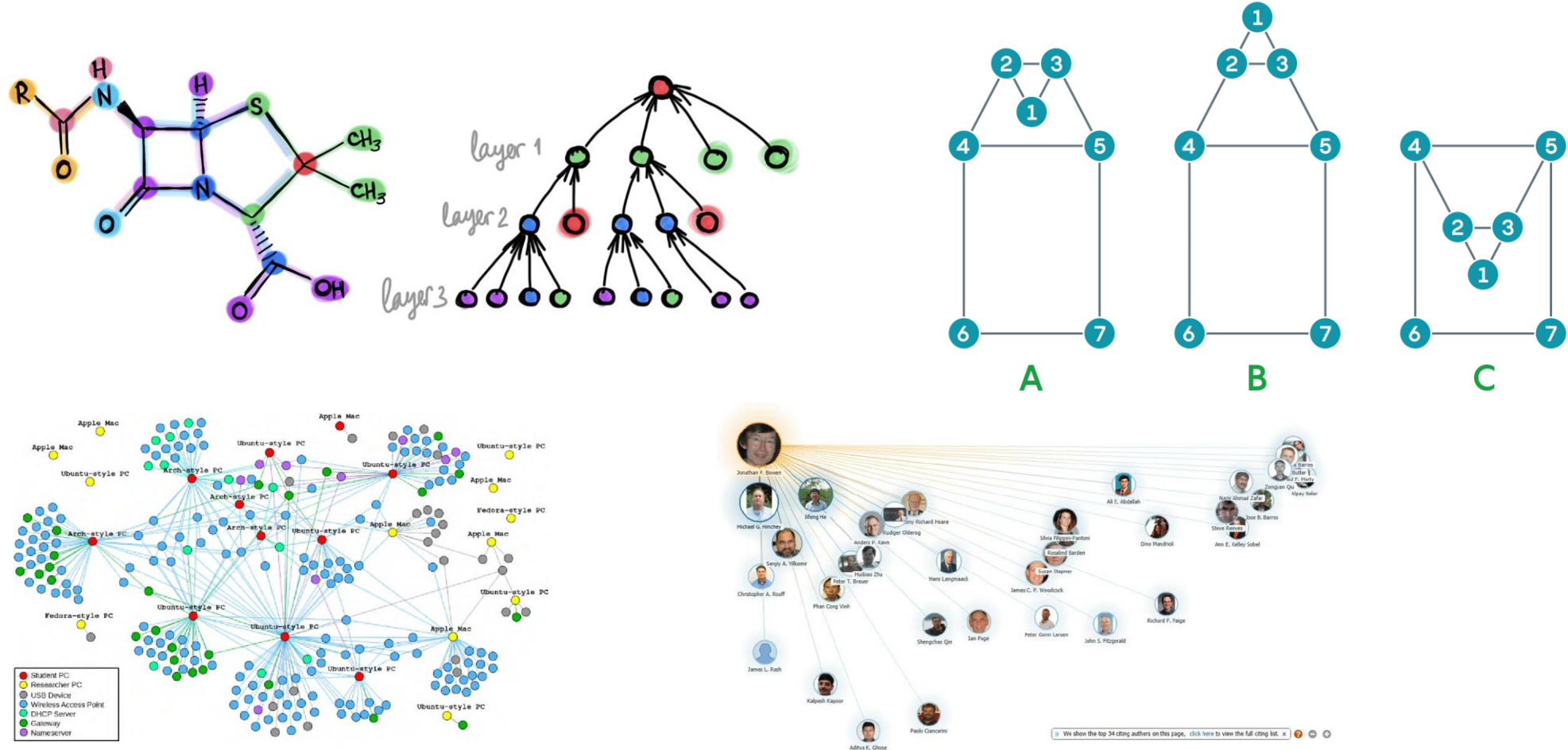
Efstratios Gavves

Lecture overview

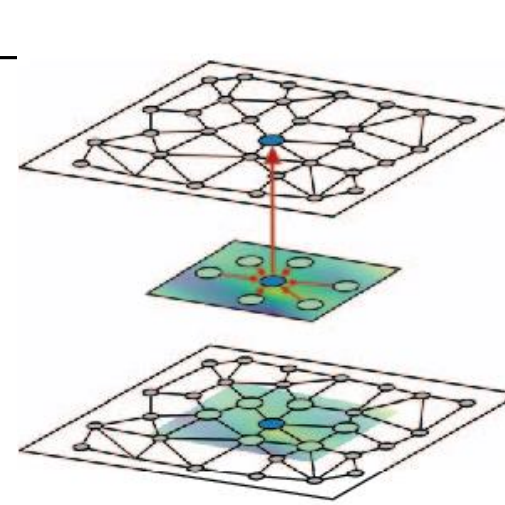
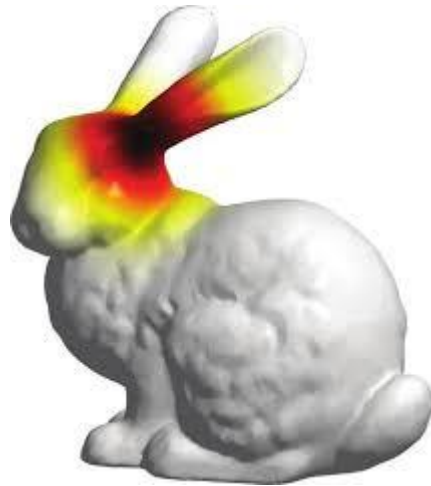
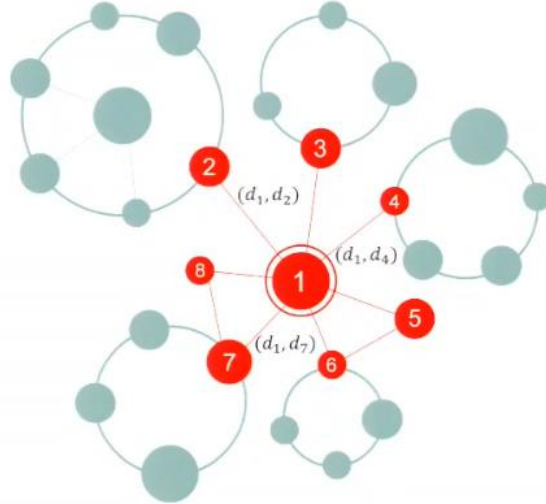
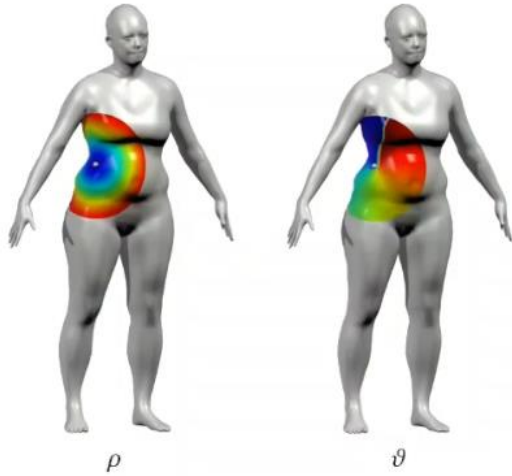
- What makes graphs special?
- Revisiting graphs
- Revisiting convolutions
- Spectral graph convolutions
- Spatial graph convolutions

What are graphs?

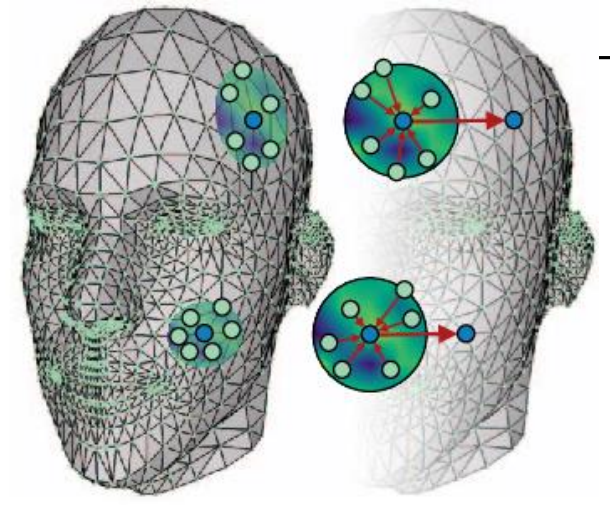
- Data structures of sets of variables (or objects) connected with each other



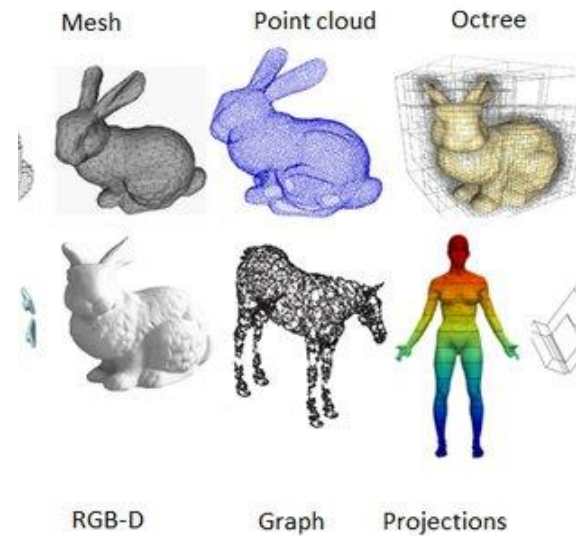
Graphs as geometry



(a) Filtering of graphs

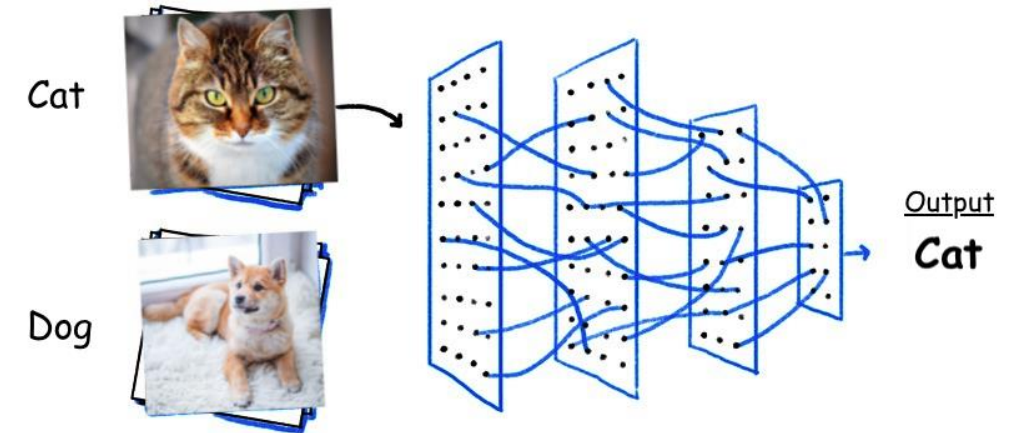
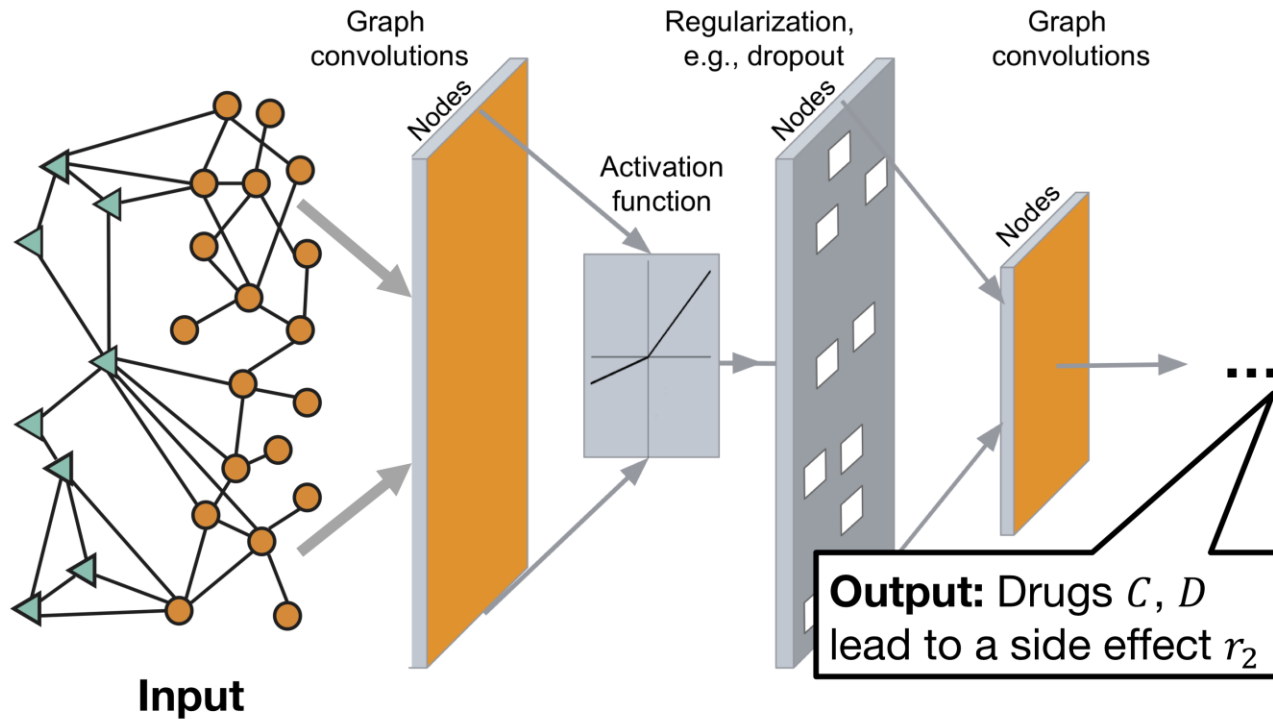


(b) Filtering of meshes



Classifying graphs

- Making predictions at the level of a graph
- Even an image can be thought as regular graph (grid)



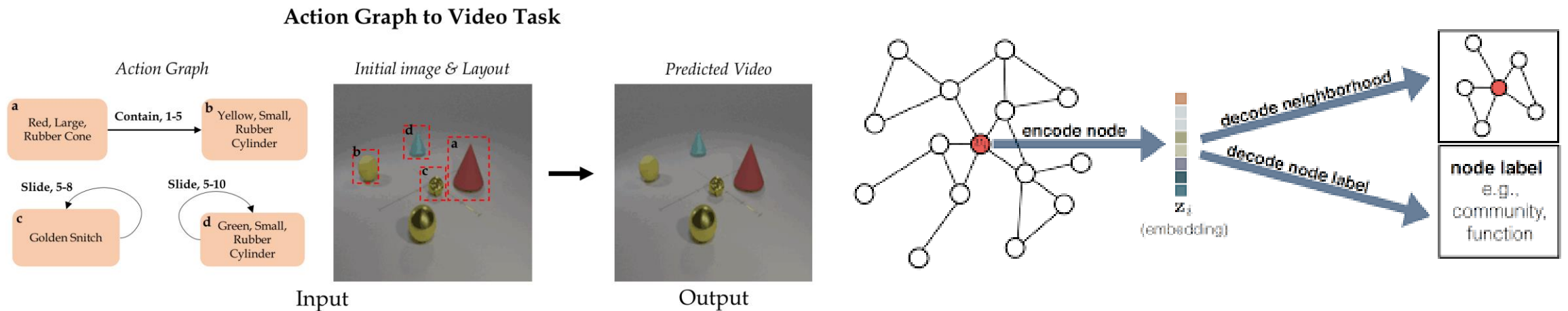
Classifying nodes

- E.g., who is likely to have coronavirus?
- Or which pixel nodes correspond to car or a pedestrian?



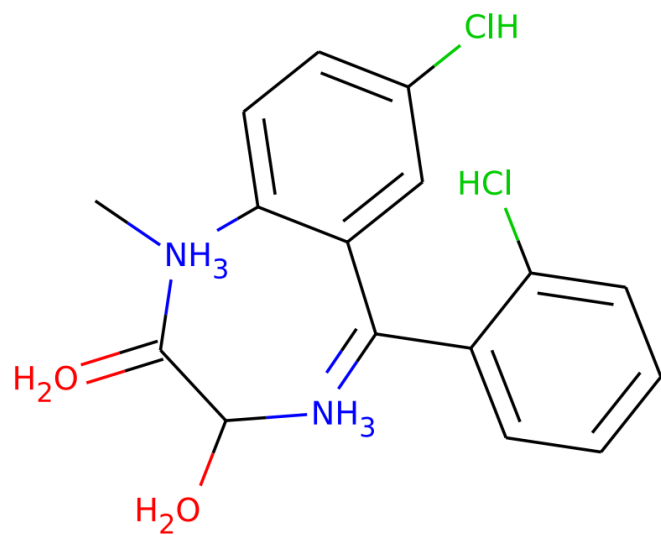
Classifying edges

- E.g., actions between objects
- Filling missing edges

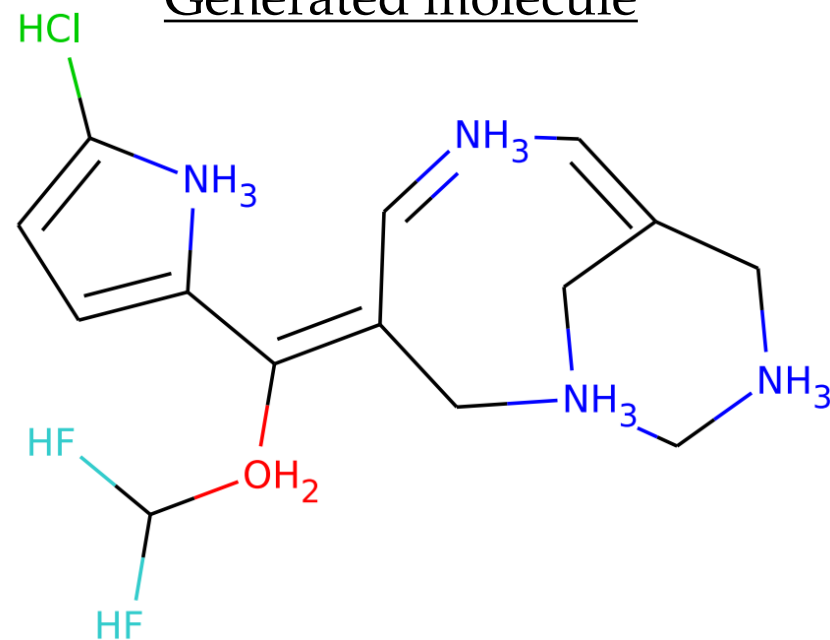


Graph generation

Example molecule



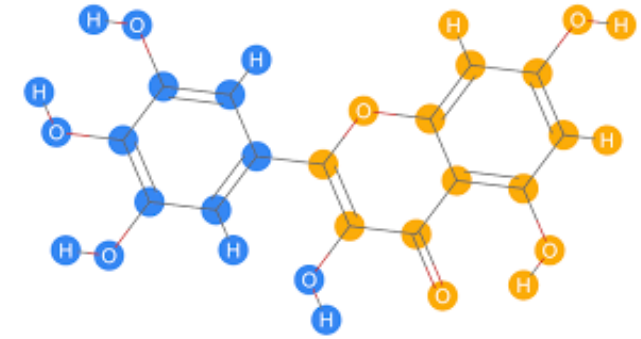
Generated molecule



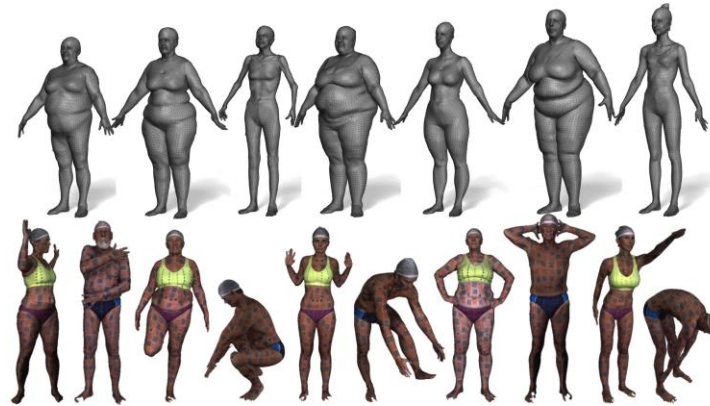
P. Lippe, E. Gavves, Categorical Normalizing Flows via Continuous Transformations, in submission to ICLR 2021

Graphs can be static, varying, or even evolving with time

- Molecule graphs do not change



- 3D mesh grids can deform



- Graph relations between objects change continuously

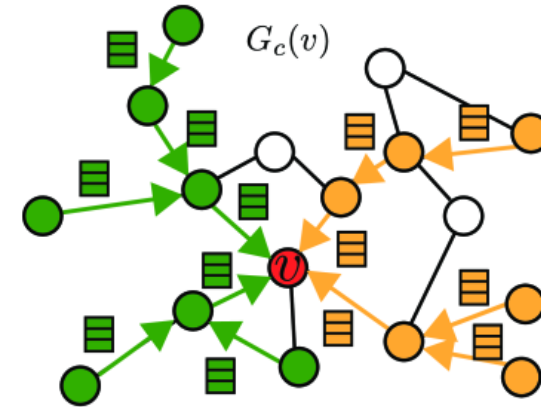


Regular structures vs graphs

- Regular structures are subsets of graphs
 - E.g., images are grid graphs



- Convolution + pooling
- Local neighborhood: fixed window
- Constant number of neighbors
- With fixed ordering
- Translation equivariance
- $O(n)$ per layer complexity



- Message passing + coarsening
- Local neighborhood: 1-hop
- Different number of neighbors
- No ordering of neighbors
- Local permutation equivariance
- $O(n)$ per layer complexity