

Today: - what if my graph grows, what if too large?
↳ graph limits: graphons

→ We understand the different ways a graph can be perturbed, and we know how to design stable GNNs.

↳ 1) but what if my graph grows?
2) or, what if it is too large and I don't have enough resources to train on it?
(recall a GNN forward pass requires $O(LK|E|)$ complexity)

For 1): can we measure how close two large graphs are? ^{b)} If we can, and if the GNN is continuous wrt this metric, then we are good. ^{d)}
what does it mean for graphs to be close?
↳ convergence to a common limit ^{a)}

For 2) can we sample a subgraph? c)

↳ random graph model

We have more questions than answers! Let's take them on one at a time.

a) What does it mean for graphs to be close?

We are statisticians/probabilists; the prevailing perspective is that we consider graphs to be close if sampled subgraphs have similar distributions, or, equivalently, if they have similar subgraph counts

↳ can be measured using homomorphism densities

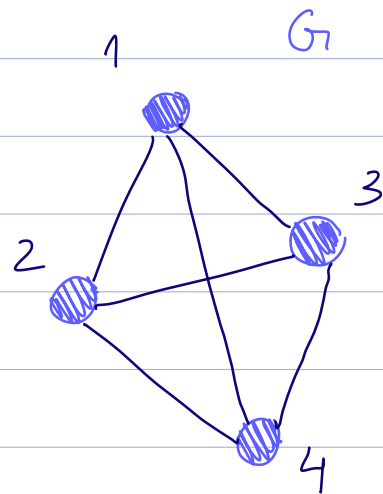
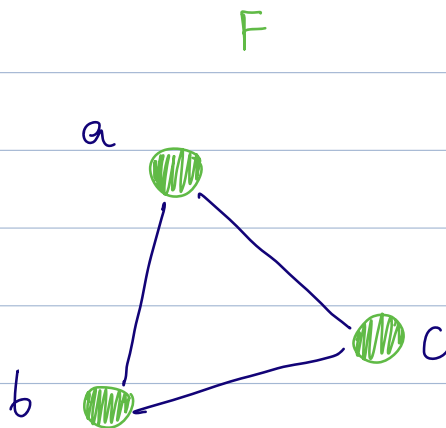
Recall from Lecture 11:

(DEF) Graph homomorphism: Let $G = (V, E)$ & $F = (V', E')$. A homomorphism from F to G is a map $\gamma : V' \rightarrow V$:

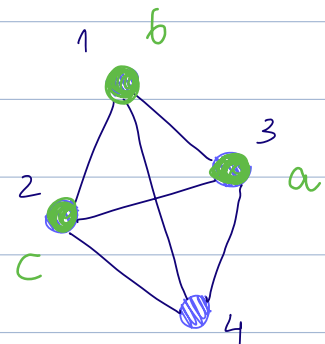
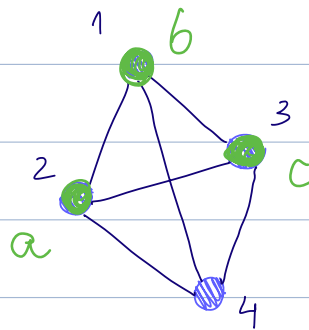
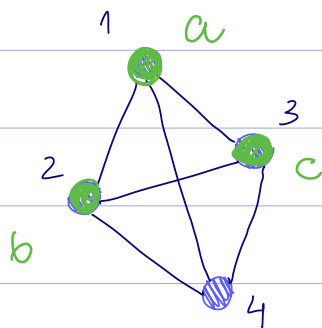
$$(i, j) \in E' \Rightarrow (\gamma(i), \gamma(j)) \in E$$

I.e., homomorphisms are adjacency preserving maps
(of F)

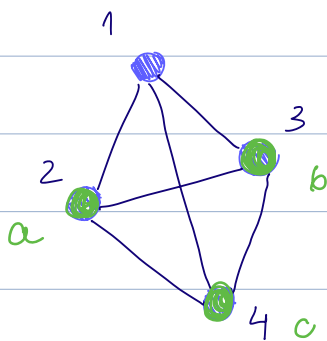
E.g.:



Homomorphisms from F to G :

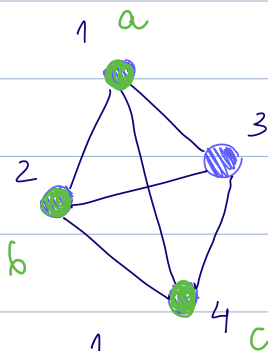


etc...

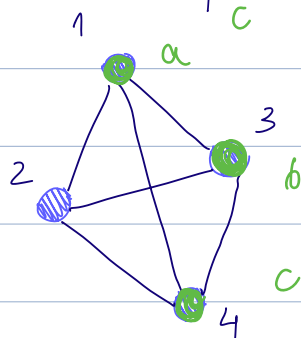


(6 in total)

(plus other permutations
of a, b, c) (6 total)



||



||

We will denote the total number of homomorphisms
from F to G $\text{hom}(F, G)$. In the example,

$$\text{hom}(F, G) = 24$$

(DEF) Homomorphism density: the hom. density from F to G ,
denoted $t(F, G)$ is:

$$t(F, G) = \frac{\text{hom}(F, G)}{|V|^{1V'}}$$

In the example, $t(F, G) = \frac{24}{4^3} = \frac{3}{8}$

\Rightarrow For now, we will say two graphs G_1 & G_2 are close if, for all motifs F (undirected, unweighted graphs, w/ one edge per node pair & no self loops), $t(F, G_1) \approx t(F, G_2)$

\rightarrow Convergent graph sequences & graphons (Lovász, Chayes, Borgs, Vesztegombi) 2008-onwards

(DEF) Let $(G_n)_n$ be a graph sequence such that $\lim_{n \rightarrow \infty} t(F, G_n)$ exists for all motifs F .

Then, this sequence is convergent and its limit is given a graphon.

(DEF) A graphon is a symmetric, bounded, measurable function $W: [0, 1]^2 \rightarrow [0, 1]$
($[0, 1] \times [0, 1]$)

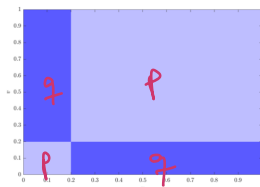
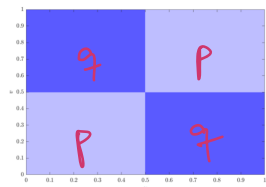
I.e., it is a bounded kernel

Obs. 1: A more general definition is $w: \Omega \times \Omega \rightarrow [0,1]$ where Ω is some sample space endowed with some probability measure f . But, since we can always map Ω into $[0,1]$ using a measure-preserving map (as long as CDF of f is strictly monotone \rightarrow exercise to verify), we'll stick with $[0,1]$ as our node sample space.

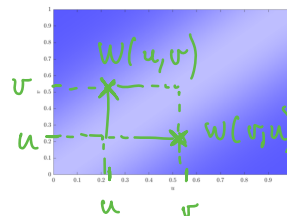
Obs. 2: The codomain of w may be $[0,B]$, $B < +\infty$, but it is typical to have $B=1$ since $w(x,y)$ often represents a probability.

The easiest way to think of a graphon is as a graph with an uncountable number of nodes $u \in [0,1]$, and edges (u,v) with weights $w(u,v)$.

E.g.



$$w(u,v) = \exp\left(-\frac{(u-v)^2}{\sigma^2}\right)$$



$$w(u,v) = w(v,u)$$

(DEF) Graphon homomorphism density

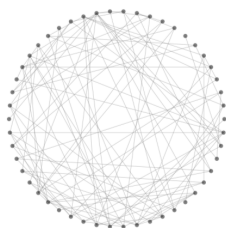
The density of homomorphisms from a graph motif $F=(V',E')$ to a graphon W is denoted $t(F,W)$ and defined as:

$$t(F,W) = \int_{[0,1]^{|V'|}} \prod_{(i,j) \in E'} W(u_i, u_j) \prod_{i \in V'} du_i.$$

If $W(u,v) < 1 \quad \forall u, v$, $t(F,W)$ can be interpreted as the probability of sampling the motif F from the graphon W .

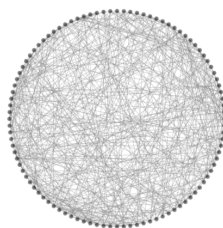
$n \rightarrow \infty$
 $t(F, G_n) \rightarrow t(F, W) \iff (G_n)_n$ is a conv. graph sequence w/ limit W
 $\forall F$

E.g.: G_{50}



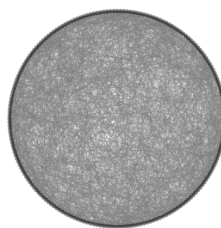
(a) 50 nodes

G_{100}



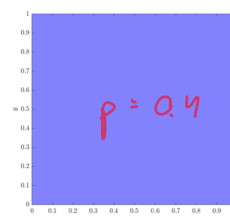
(b) 100 nodes

G_{200}



(c) 200 nodes

W



(d) Graphon

This notion of convergence is called left convergence as it deals with left homomorphisms $t(F, G_n)$, $t(F, w)$

This is not the only way to define/identify (dense) convergent graph sequences. Another definition based on the convergence of min-cuts (micro-canonical ground state energy in physics) exists, based on right homomorphisms $(t(G_n, F), t(w, F))$. For dense graphs, left & right convergence are equivalent.

b) Can we measure how close two large graphs are?

While $t(F, G)$ gives us a way to define convergence, it is difficult to use it to measure the distance between graphs, as we'd have to compute $t(F, G_1)$ & $t(F, G_2)$ for all motifs F .

► Cut distance for graphs

① Graphs G & G' with same n & same node sets V, V' (i.e., same node labeling)

We define two distances:

L_1 norm or edit distance: $d_1(G, G') = \|A - A'\|_1$,
where A, A' are the (unweighted) adjacencies of G, G'

Cut distance: $d_{\square} = \|A - A'\|_{\square}$,

where $\|\cdot\|_{\square}$ is the cut norm:

$$\|B\|_{\square} = \max_{S \subseteq [n], T \subseteq [n]} \left| \sum_{s \in S, t \in T} B_{ts} \right|$$

I.e., given a (graph) matrix B , $\|B\|_{\square}$ is "the size" (total number of edges, or sum of edge weights) of the max cut.

② Graphs G & G' with the same number of nodes:

$$\text{Cut distance: } \hat{S}_\square(G, G') = \min_{P \in \mathcal{P}} \|A - P^T A' P\|_\square$$

I.e., we must look at all possible permutations of the nodes of G' (while fixing the node labels in G)

③ Graphs G_n & G_m with \neq number of nodes ($n \neq m$)

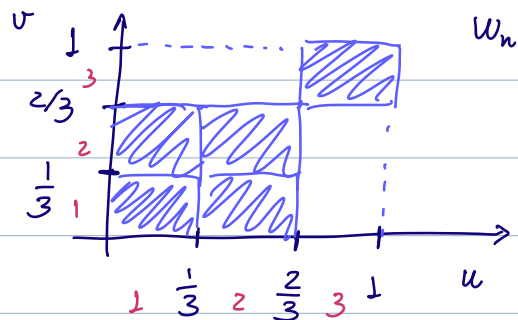
$$\text{Define } W_n(u, v) = \sum_{i=1}^n \sum_{j=1}^n A_{ij} \mathcal{I}(u \in I_i) \mathcal{I}(v \in I_j)$$

where \mathcal{I} is the indicator function and I_j is defined as:

$$I_j = \begin{cases} \left[\frac{j-1}{n}, \frac{j}{n} \right], & 1 \leq j \leq n-1 \\ \left[\frac{n-1}{n}, 1 \right], & j = n \end{cases}$$

E.g.:

$$A_n = \begin{matrix} & \begin{matrix} 1 & 2 & 3 \end{matrix} \\ \begin{matrix} 1 \\ 2 \\ 3 \end{matrix} & \begin{bmatrix} 1 & 1 & 0 \\ 1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \end{matrix}$$



w_n is called the graphon induced by G_n .

Back to the cut distance:

Using transformation $(*)$, we can represent both G_n and G_m as kernels on $[0,1]^2$, so they are compatible objects. But now we need a notion of cut norm for kernels.

Cut norm: Let w be a kernel in $[0,1]^2$.

Its cut norm is defined as:

$$\|w\|_{\square} = \sup_{S, T \subseteq [0,1]} \left| \int \int_S w(u,v) du dv \right|$$

We're not done yet! We need to take into account "node relabelings" (permutations of I_1, I_2, \dots).

So for two kernels (graphons), we define the cut metric:

$$\delta_{\square}(w, w') = \inf_{\phi} \|w^{\phi} - w'\|_{\square}$$

where $w^{\phi}(u, v) = w(\phi(u), \phi(v))$ and ϕ are measure-preserving bijections.