

So for two Kernels (graphons), we define the cut metric: SD (W,W') = inf || W - W'|| where  $w^{\phi}(u,v) = W(\phi(u),\phi(v))$  and  $\phi$ measure-preservina bijections Hence So (Gn, Gm) = So (Wn, Wm) Sequences of graphs (Gn)n converging to a graphon W also converge in the cut  $G_n \xrightarrow{n \to \infty} W \qquad \stackrel{(=)}{\longrightarrow} \|W_n - W\|_{\mathbb{Q}} \xrightarrow{n \to \infty} O$ 

In fact, the way to show that left & right convergence are equivalent is by showing that they are both equivalent to convergence in the cut metric.

For proofs, check the book "Louge networks & convergent
For proofs, check the book "Lange networks & convergent araph sequences" by Lovaisz (available anline).  (For = w/ hom. density carv., check counting
For ≡ w/ hom. density conv., check counting
à inverse counting lemmas)
- Relationship between the cut norm & Lp
norms for graphons (and graphon distances (codomain [-1,1])
(codomain [-1,1])
Let $W: (0,1)^2 \rightarrow (-1,1]$
Trivial inequalities:    w   =      w   =    w   =      w   =
=> eonvergence in Lp, p>1, implies cut norm
=> eonvergence in Lp, p>1, implies cut norm convergence.
con vergence.
=> eonvergence in Lp, p>1, implies cut norm convergence.  In the other direction, we have:
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con vergence.
In the other direction, we have: $\  \mathbf{w} \ _{\mathbf{a},\mathbf{a}} \leq \sqrt{4 \  \mathbf{w} \ _{\mathbf{b} \to 1}} \leq \sqrt{16} \  \mathbf{w} \ _{\mathbf{D}}$
In the other direction, we have:

Pf: We start by noting that

$$\|W\|_{D} = \sup_{S,T \subseteq Q(I)} \int W(u,v) du dv$$

$$= \sup_{\{g\} : Q(I), > Q(I)\}} \int W(u,v) f(u)g(v) du dv$$

$$= \int W(u,v) f(u) du$$

Now, IWII so is defined as:

Rewriting this expression as:

$$\|W\|_{\infty \to 1} = \sup_{0 \le f_1 f' \le 1} \langle T_w(g - g'), f - f' \rangle$$
 $0 \le f_1 f' \le 1$ 
 $0 \le g_1 g' \le 1$ 

We get:

$$||W||_{\infty \to 1} \leq mp \qquad \langle Twg, f \rangle$$

$$0 \leq g, f \leq 1$$

For the first, use the Riesz-Thorin interpolation theorem for complex Lp spaces:

where 
$$\theta = \min\left(1 - \frac{1}{9}, \frac{1}{9}\right)$$
,  $\rho_0, q_0 \in C1, \infty C$ 

$$W/\frac{1}{P} = \frac{1-\Theta}{P^{\circ}}$$
  $\frac{1}{9} = \frac{1-\Theta}{(1-\Theta)} \left(\frac{1-1}{9^{\circ}}\right)$ 

and 
$$p_1 = \infty$$
,  $q_1 = 1$ .

Define:

For complex functions, we thus have:

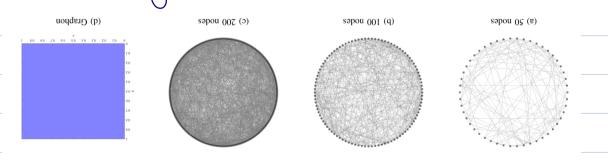
## $\|W\|_{\infty \to 1} = \|W\|_{D,C} \le 2 \|W\|_{\infty \to 1}$ $\forall$ for complex functions

Noting that  $\|W\|_{\rho \to q_0} < \|W\|_{1 \to \infty} < \|W\|_{\infty} < 1$  (since  $W \le 1$ ) campletes the proof.

3) Can we use graphons to sample subgraphs?

Yes! They are generative models as well.

In fact, all of the graphs below were sampled from the graphon.



## How to sample?

•) "Template" graphs

The simplest way to sample a graph from W is by partitioning (0,1) in a grid (regular partition):

I, U I2 V... U In

where  $I_{j} = \begin{cases} \frac{j-1}{n}, \frac{j}{n} \\ \frac{(n-1)}{n}, \frac{1}{j} \end{cases}, j = n$ 

and the node labels are  $u_j = j-1$  tj

We then define the template graph on via its adjacency,

[An] = W (mi, uj)

i.e., it is a weighted graph.

## ·) Random weighted graphs

Another type of graph we can sample from W are graphs with random nodes, where the u; are sampled randomly from (9,1), typically, uniformly, i.e.:

uj ~ Uniform CO, 13

The edges are then defined in the same way as for template graphs:

(An] = W(ui, uj) i.e., it is a weighted graph

•) "Fully" random graphs (a.K.a. W-random graphs)

These are graphs with both random nodes & edges. The node labels are once again sampled as: \(\mu\_i \sim \text{Uniform((0,1))}\)

and the edges as:
(An ]; ~ Bernoulli (w(ui, uj))
The edges are unweighted and undirected.
All of the graphs above converge to w in
some sense
- Template: trivial. Convergence in Lz implies 11.110
- Template: trivial. Convergence in Lz implies 11.110 convergence (deterministic)
- Weighted: sampling lemma (1):
w.p. at least $1-exp\left(\frac{-n}{2\log n}\right)$
$80 (Gn, W) \leq 20/\sqrt{n}$
- Random: sampling lemma 2:
W. p. at 18051 (- exp (2/ogn))
So (Gn, W) < 22 / Ylogn We'll look at the proof later.
We'll look at the proof later.

d) GNN continuity, i.e., GNN convergence?  This question has a lawy answer.  To start to answer it, we need to introduce graphon signals
This guestion has a large answer.
To start to answer it, we need to intro-
duce a raphon signals
Graphon Signal processina
A graphon signal is défined as a function
$\mathcal{E}: C_{0}, C_{1} \rightarrow \mathcal{R} \qquad (x \in \mathcal{R}^{n})$
We focus on signals in La, Z E Lz (Co, 17):
l .
$\int  \mathcal{X}(u) ^2 du \leq B \leq +\infty$
"finite energy signals"
Like a graphon, graphon signals are limits of convergent
Like a graphon, graphon signals are limits of convergent sequences of graph signals
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Induced graphon signals: Let (Gn, xn) be a graph signal. The induced graphon signal is defined as:

$$\mathcal{Z}_{n}(u) = \sum_{j=1}^{n} (x_{n})_{j} f(u \in I_{j})$$

$$f$$
 is indicator for  $j$   $=$   $\int \left(\frac{x^{-1}}{n}, \frac{1}{n}\right), 1 \le j \le n-1$ 

$$\left(\frac{n-1}{n}, \frac{1}{n}, \frac{1}{n}\right), j = n$$

E.a.

$$A_{n} = \begin{pmatrix} 1 & 1 & 0 \\ 1 & 1 & 0 \\ 2 & 3 & 4 \end{pmatrix}$$

$$A_{n} = \begin{pmatrix} 1 & 1 & 0 \\ 1 & 1 & 0 \\ 3 & 2 & 1 \end{pmatrix}$$

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$$x_n = \begin{bmatrix} 0.5 \\ 1 \\ 2 \\ 0.5 \end{bmatrix}$$
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