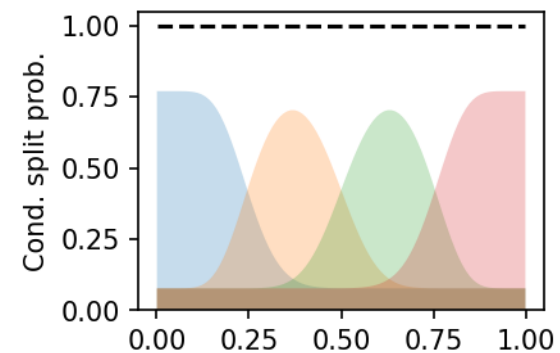
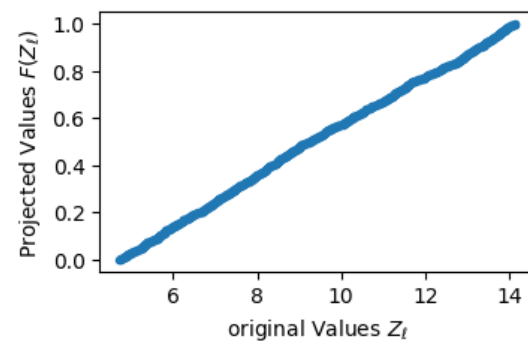
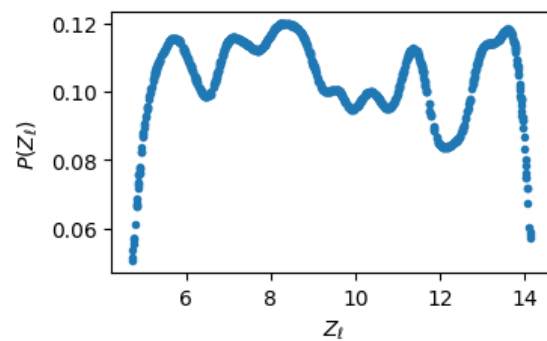
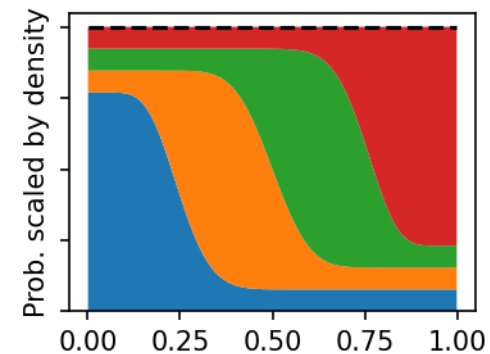


Compute $P(Z_t)$ and $F(Z_t)$

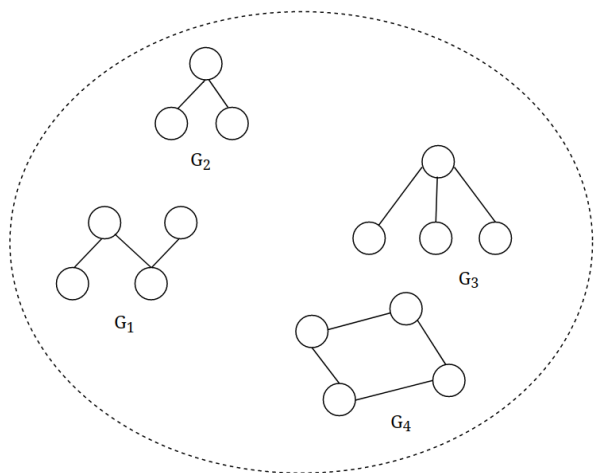
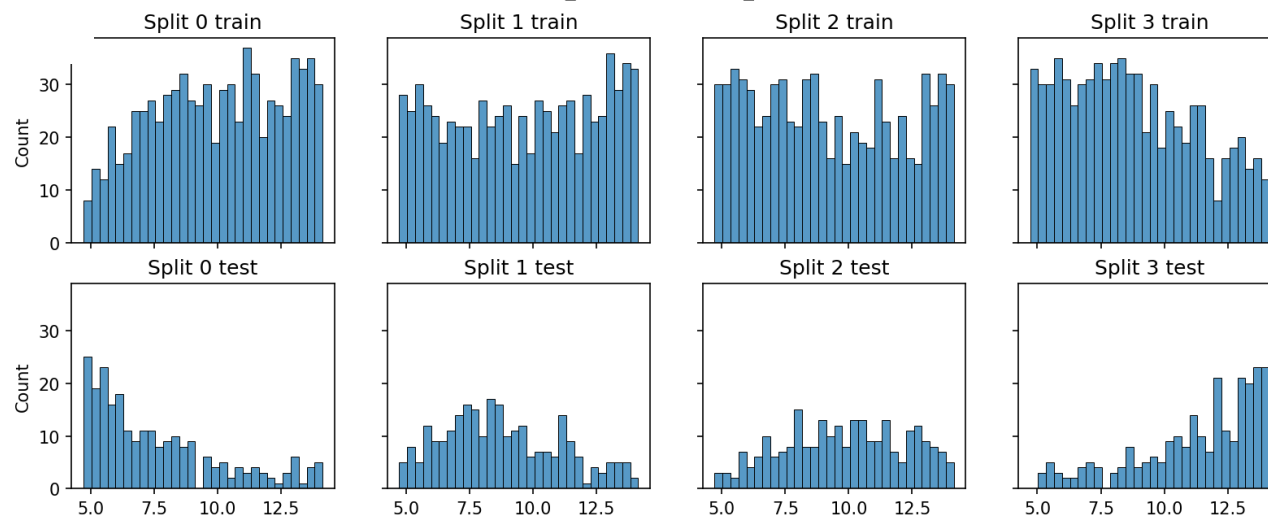


Compute conditionals

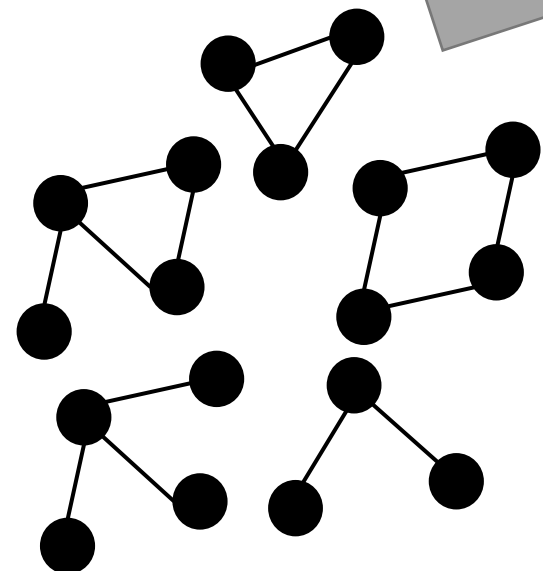


Split the data based on conditionals

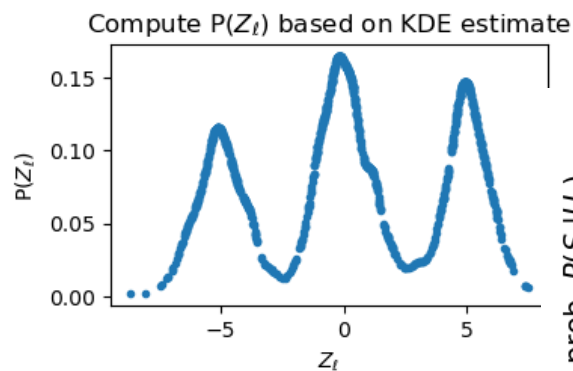
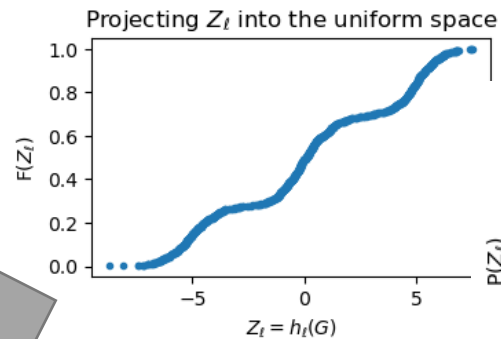
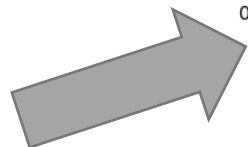
sharpness_scale=1, epsilon_base=0.1



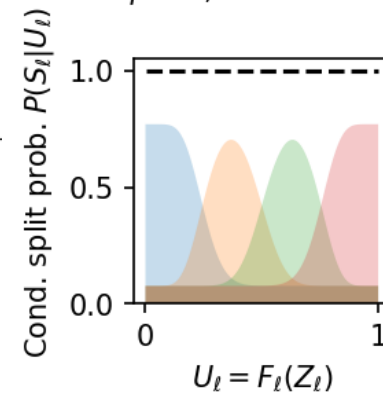
Dataset of
Graphs



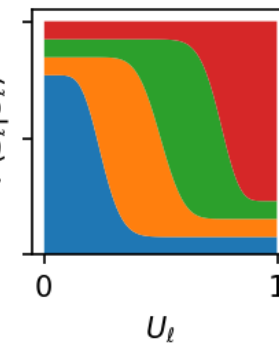
Dataset of Graphs



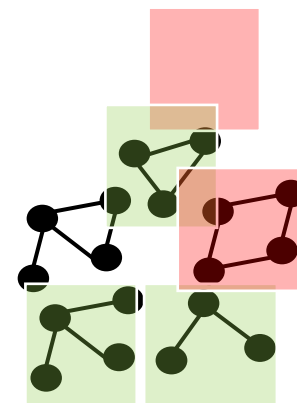
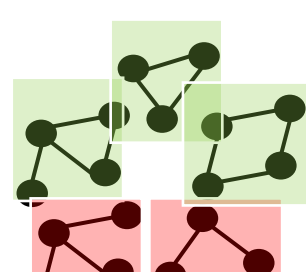
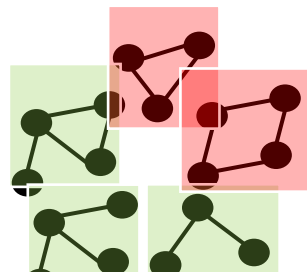
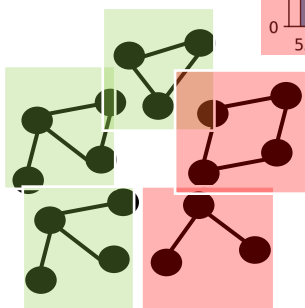
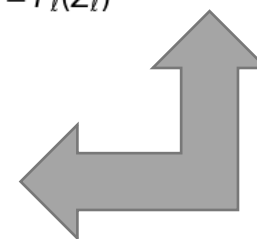
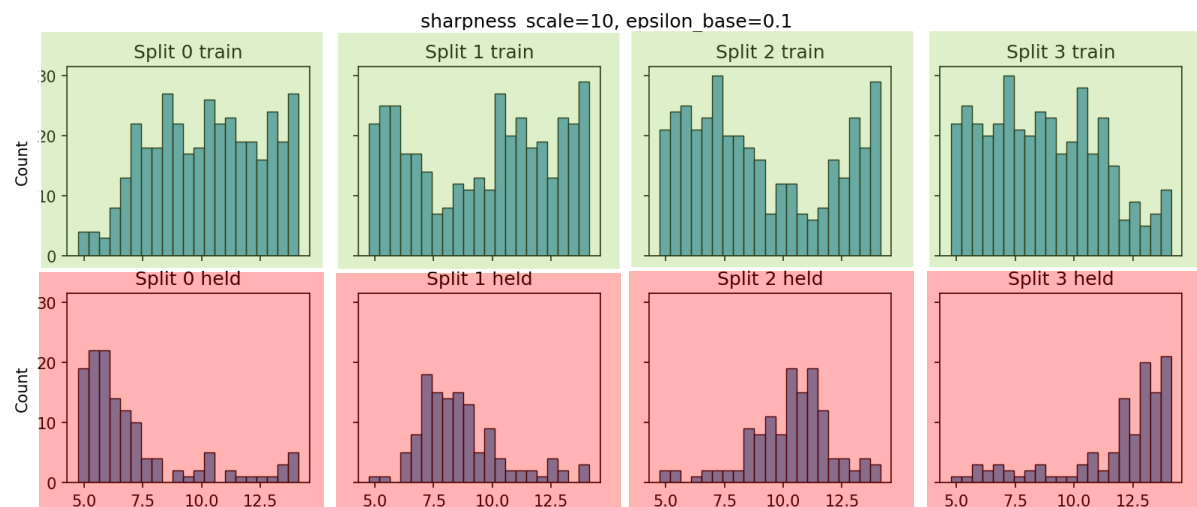
Compute conditionals using Beta distributions with $\psi=10$, and mix in uniform with factor $\epsilon_b=0.1$

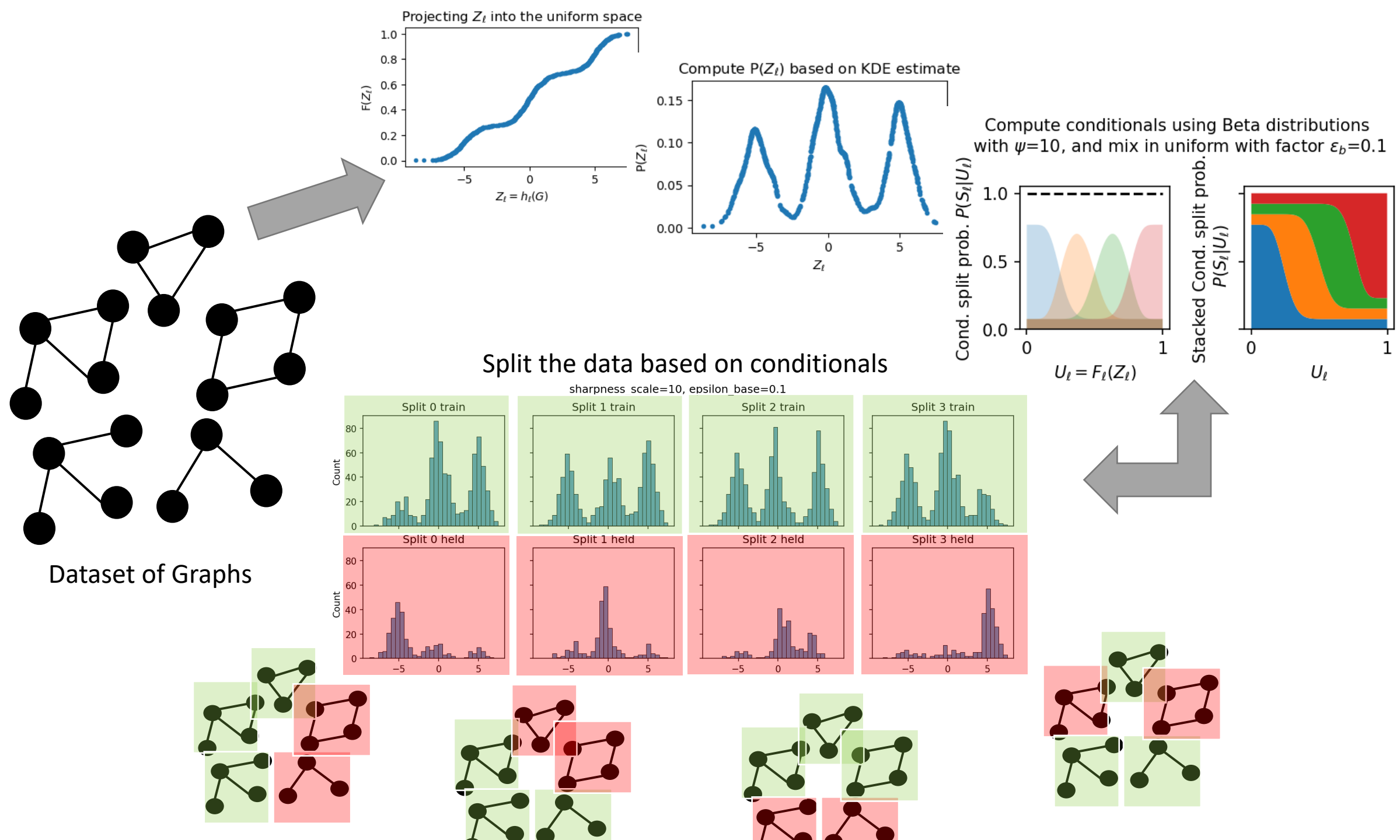


Stacked Cond. split prob. $P(S_l|U_l)$

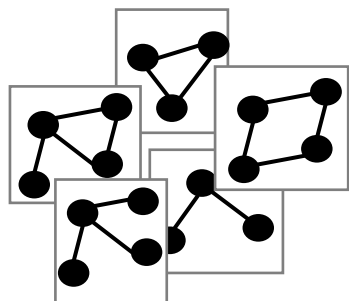


Split the data based on conditionals



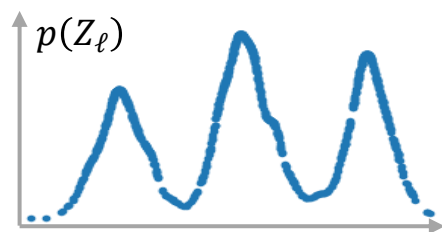


Dataset of Graphs



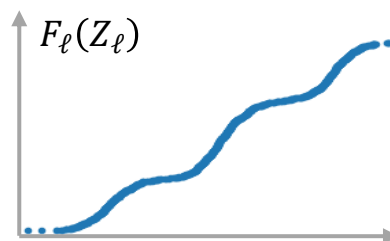
$$\mathcal{G} = \{G_i\}_{i=1}^{|\mathcal{G}|}$$

1. Compute Graph Properties



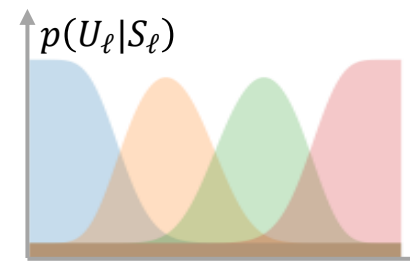
$$Z_\ell = h_\ell(G)$$

2. Project to unit interval via CDF



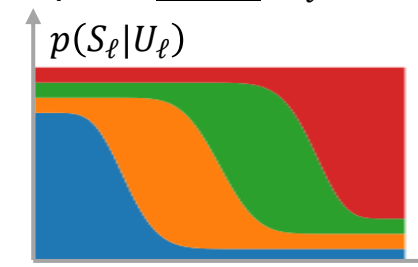
$$Z_\ell = h_\ell(G)$$

3. Define split distributions on $[0,1]$



$$U_\ell = F_\ell(Z_\ell)$$

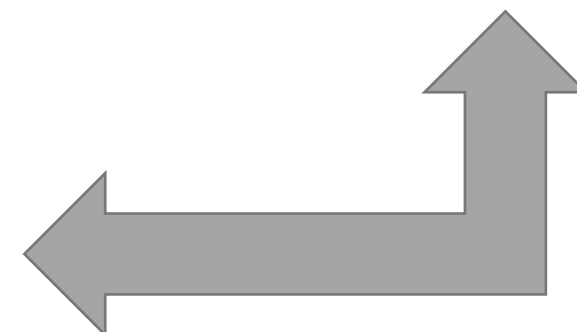
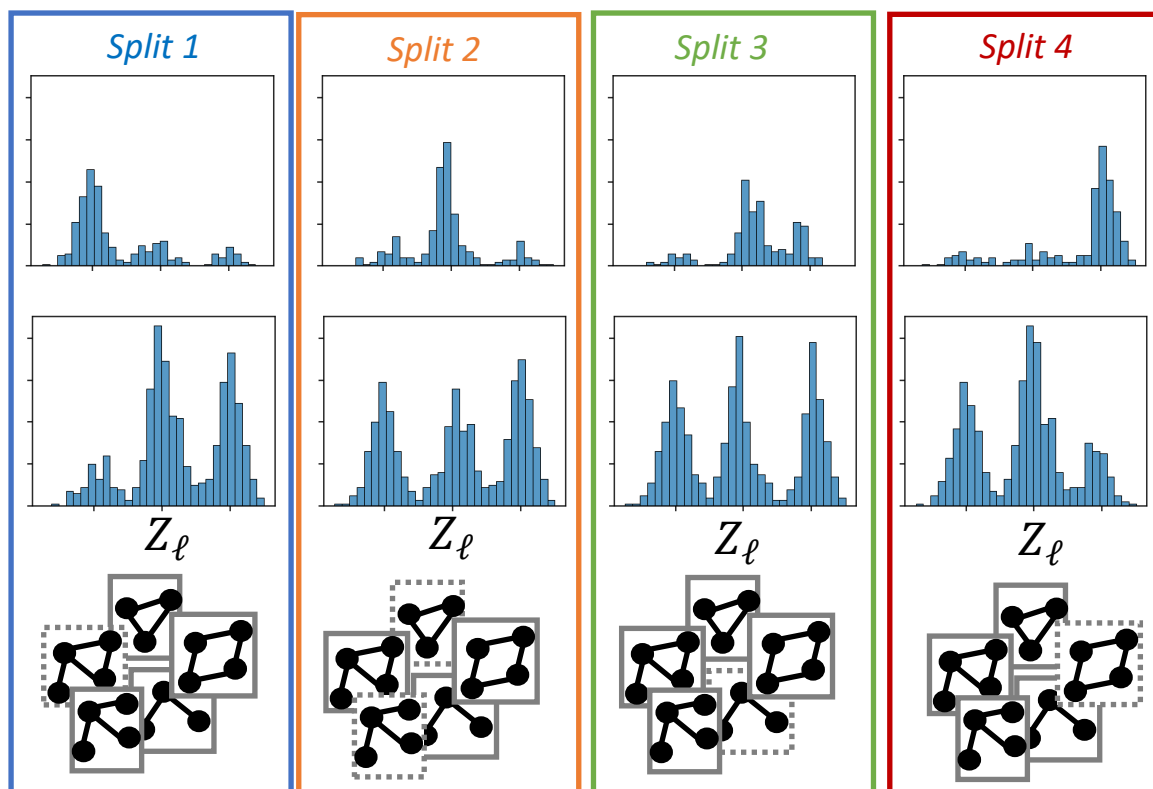
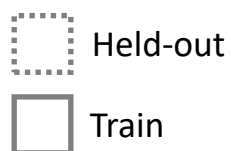
4. Compute split prob given U_ℓ



$$U_\ell = F(Z_\ell)$$

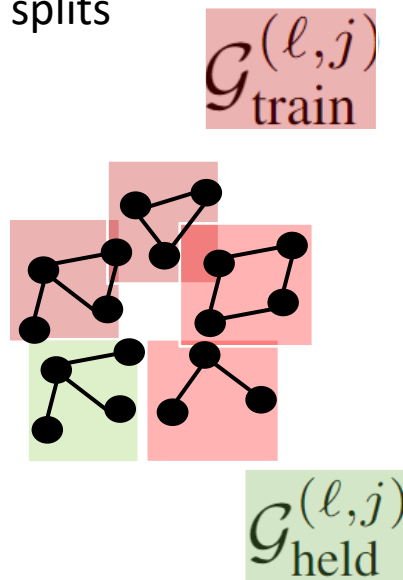
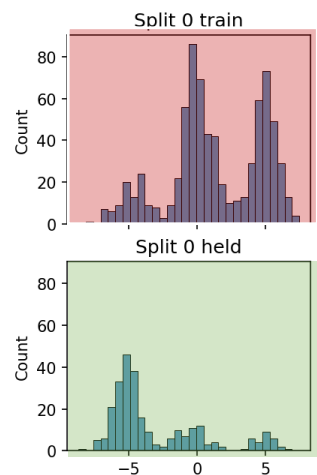
Held-Out
 $S_\ell = j$

Train
 $S_\ell \neq j$



5. Sample $S_\ell \sim p(S_\ell | U_\ell)$
for each graph

Focusing on one of the splits

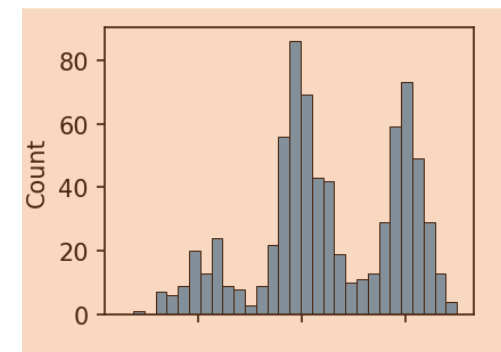


Use to train
model

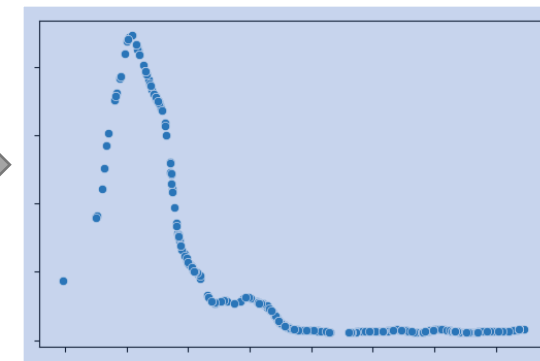
Model

Model
generates

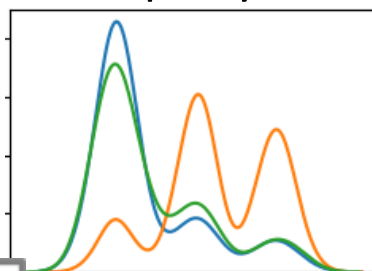
$\mathcal{G}_{\text{gen}}^{(\ell,j)}$



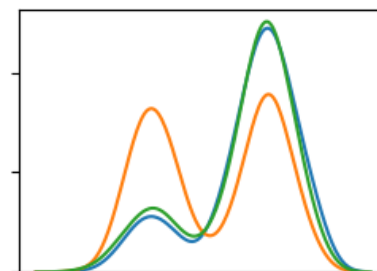
Calculated weights



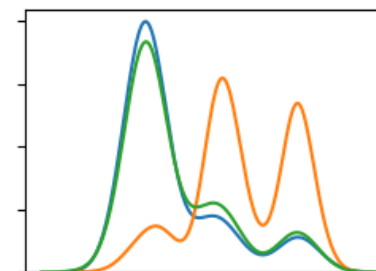
Property 1



Property 2



Property 3



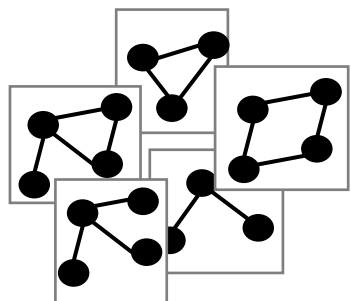
Finally calculate KS statistic between
reweighted $\mathcal{G}_{\text{gen}}^{(\ell,j)}$ and $\mathcal{G}_{\text{held}}^{(\ell,j)}$

Use weights to
reweight samples
of $\mathcal{G}_{\text{gen}}^{(\ell,j)}$ for all
the dataset
properties

— \mathcal{G}_{gen} Reweighted
— \mathcal{G}_{gen} Unweighted
— $\mathcal{G}_{\text{held}}$

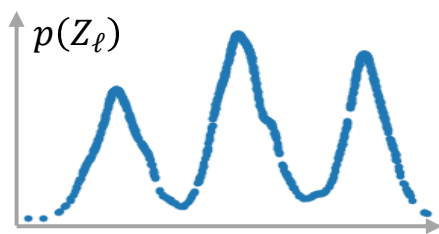
— $\mathcal{G}_{\text{gen}}^{(\ell,j)}$ Reweighted
— $\mathcal{G}_{\text{gen}}^{(\ell,j)}$ Unweighted
— $\mathcal{G}_{\text{held}}^{(\ell,j)}$

Dataset of Graphs



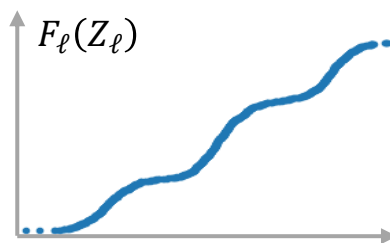
$$\mathcal{G} = \{G_i\}_{i=1}^{|\mathcal{G}|}$$

1. Compute Graph Properties



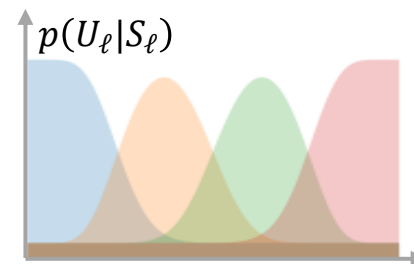
$$Z_\ell = h_\ell(G)$$

2. Project to unit interval via CDF



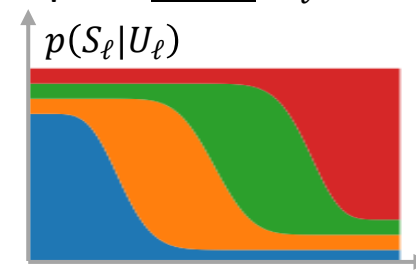
$$Z_\ell = h_\ell(G)$$

3. Define split distributions on $[0,1]$



$$U_\ell = F_\ell(Z_\ell)$$

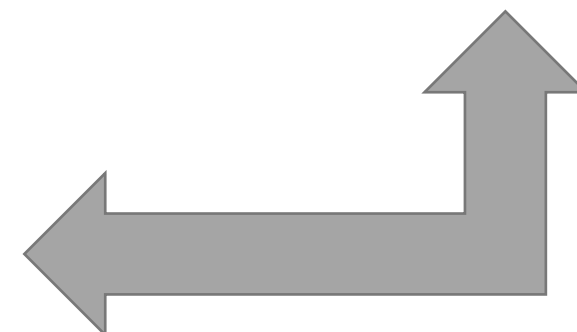
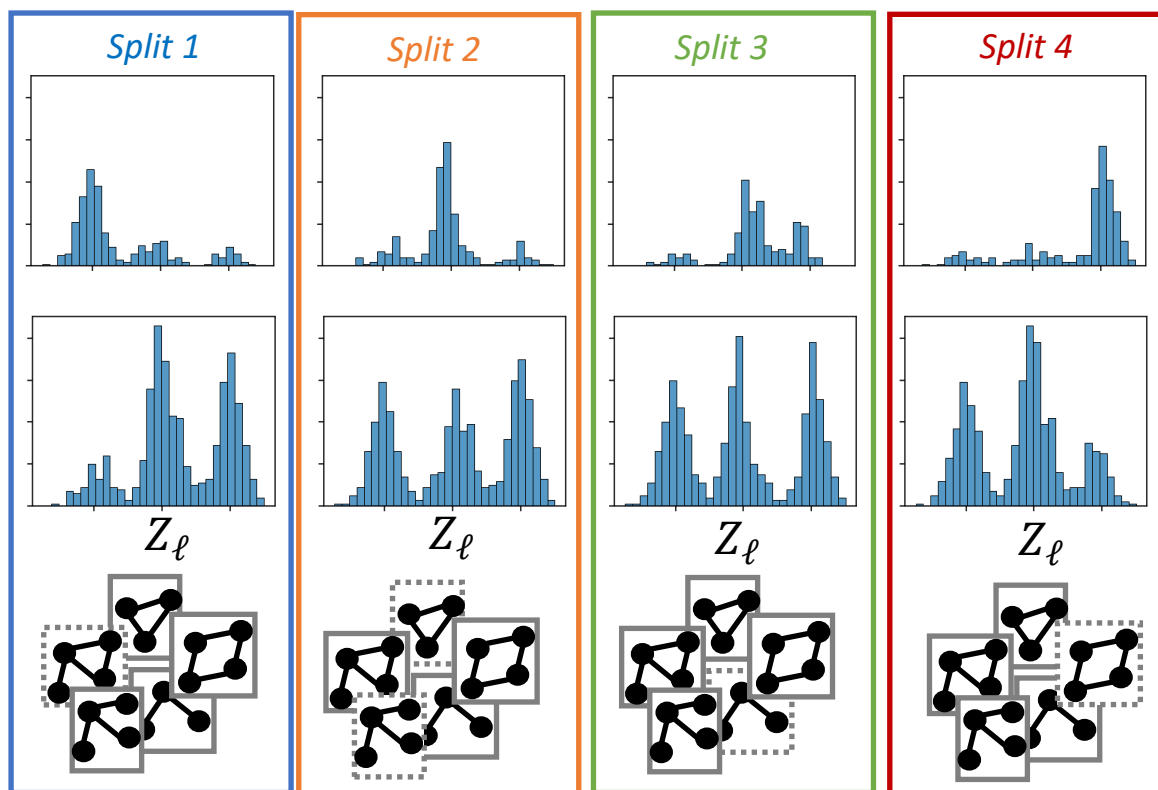
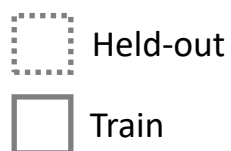
4. Compute split prob given U_ℓ



$$U_\ell = F(Z_\ell)$$

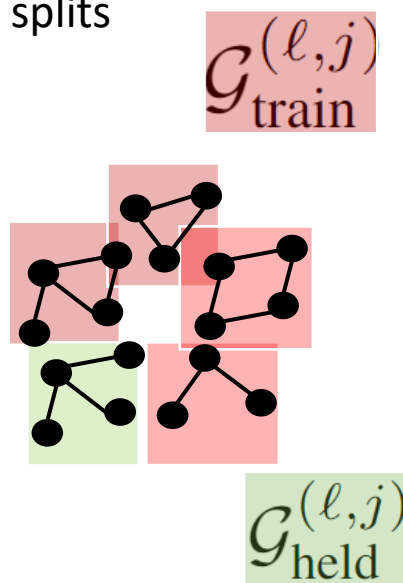
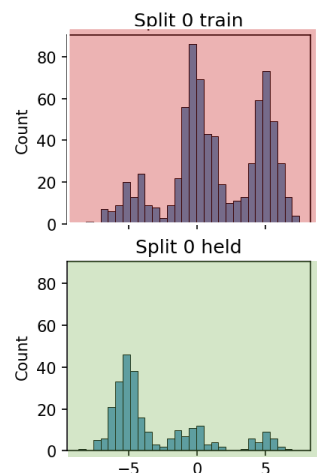
Held-Out
 $S_{i,\ell} = j$

Train
 $S_{i,\ell} \neq j$



5. Sample $S_{i,\ell} \sim p(S_{i,\ell} | U_{i,\ell})$ for each graph

Focusing on one of the splits

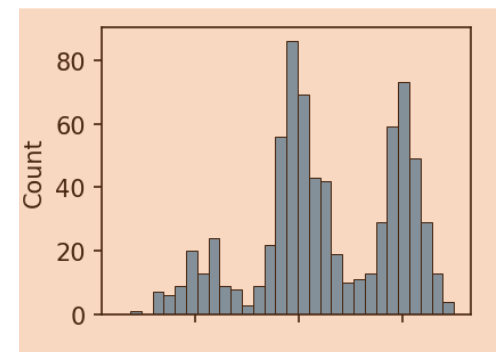


Use to train
model

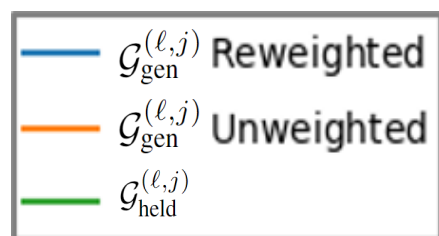
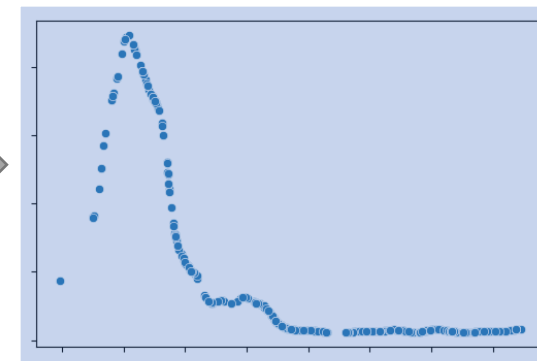
Model

Model
generates

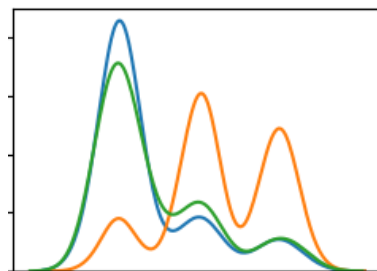
$\mathcal{G}_{\text{gen}}^{(\ell,j)}$



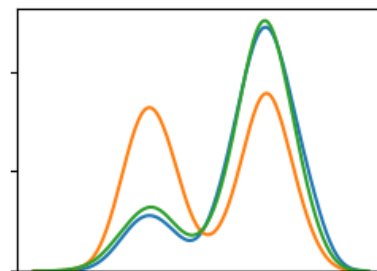
Calculated weights



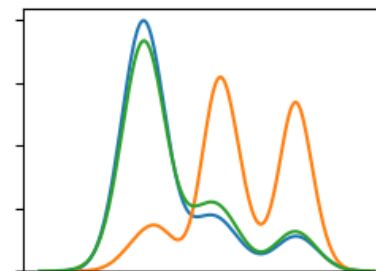
Property 1



Property 2

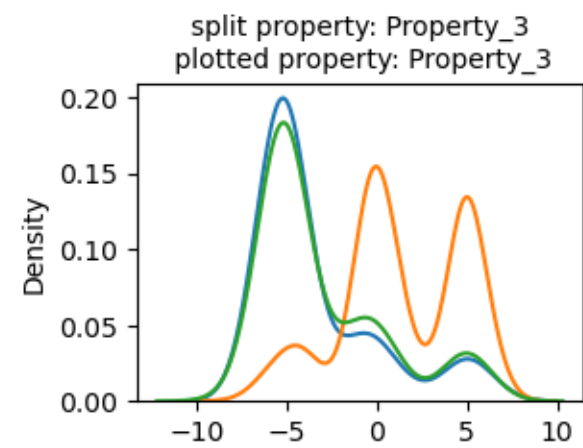
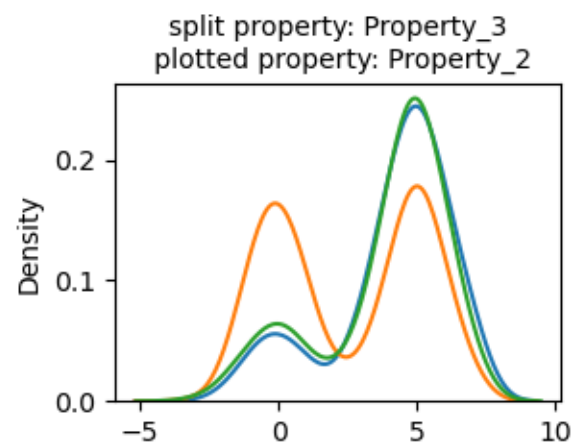
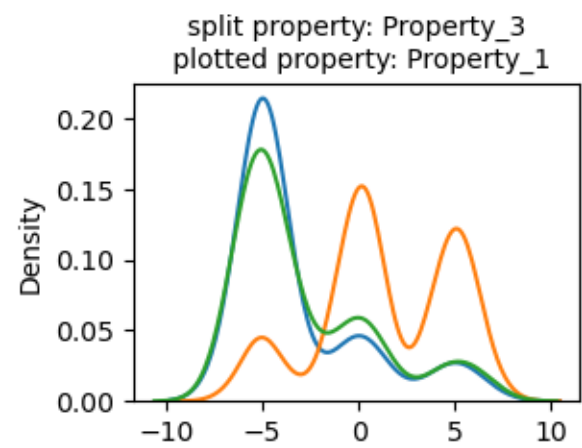


Property 3

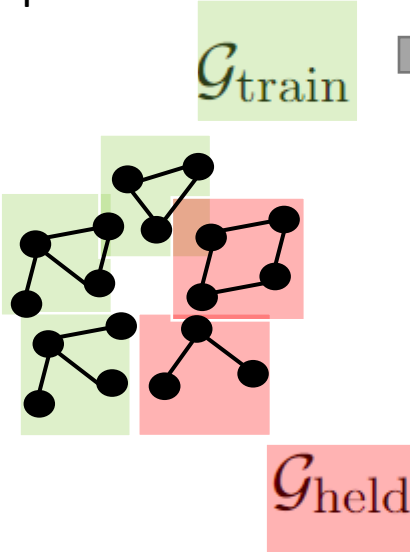
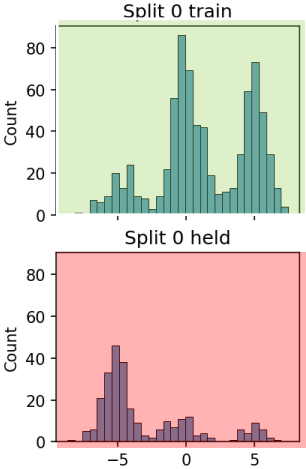


Finally calculate KS statistic between
reweighted $\mathcal{G}_{\text{gen}}^{(\ell,j)}$ and $\mathcal{G}_{\text{held}}^{(\ell,j)}$

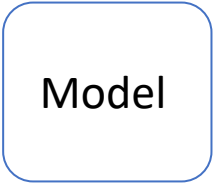
Use weights to
reweight samples
of $\mathcal{G}_{\text{gen}}^{(\ell,j)}$ for all
the dataset
properties



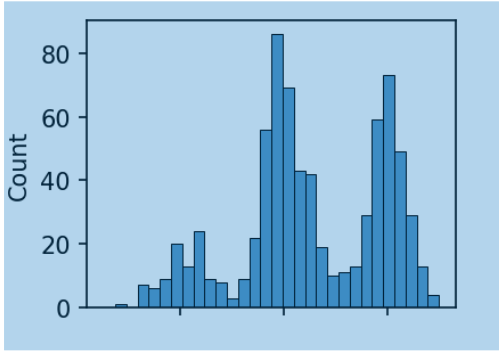
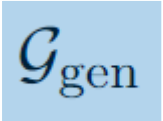
Focusing on one of the splits



Use to train
model

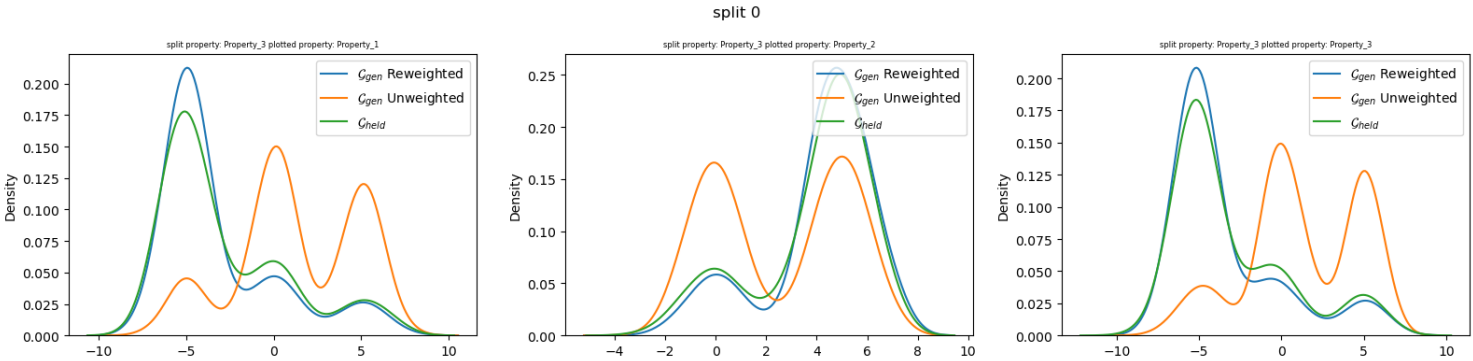
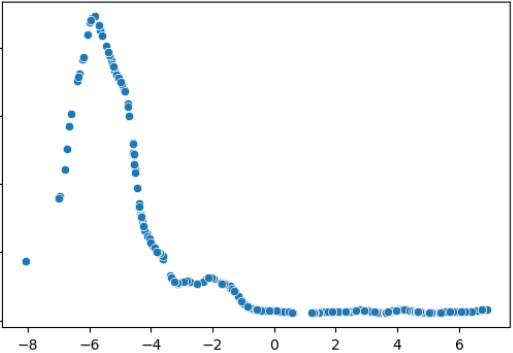


Model
generates



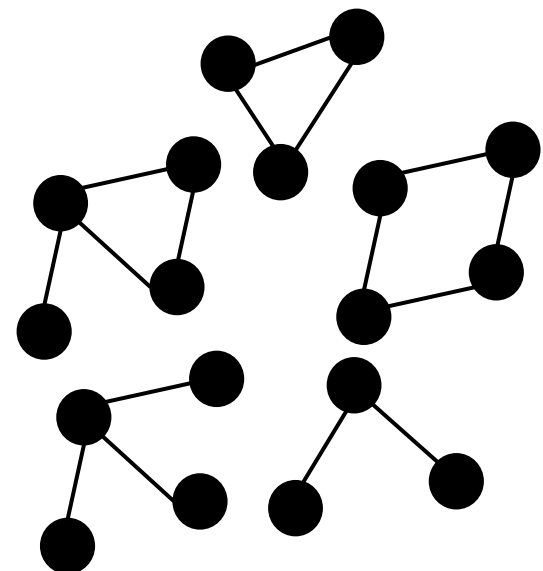
Use to Calculate
Weights to reweight G_{gen}

Calculated weights

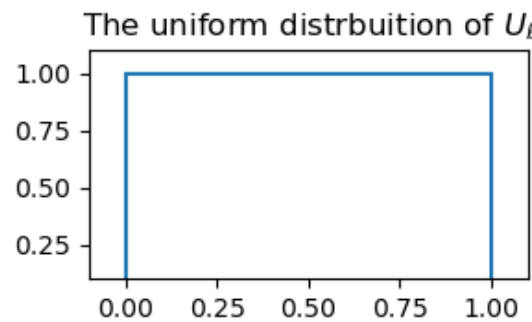
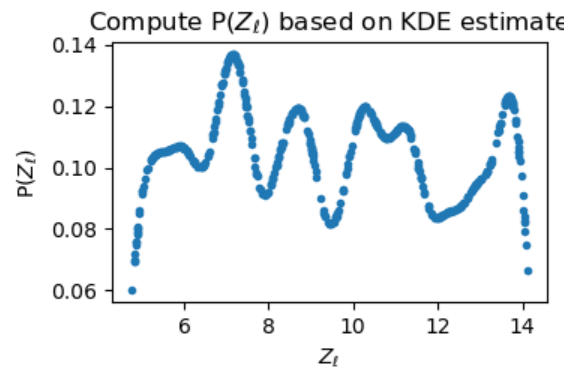
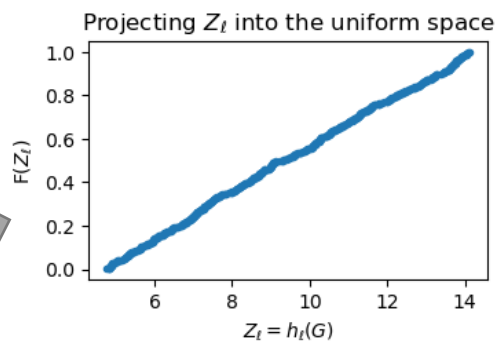
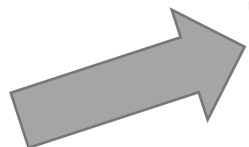


Finally calculate KS statistic between
reweighted G_{gen} and G_{held}

Use weights to
reweight samples
of G_{gen}
for all the dataset
properties

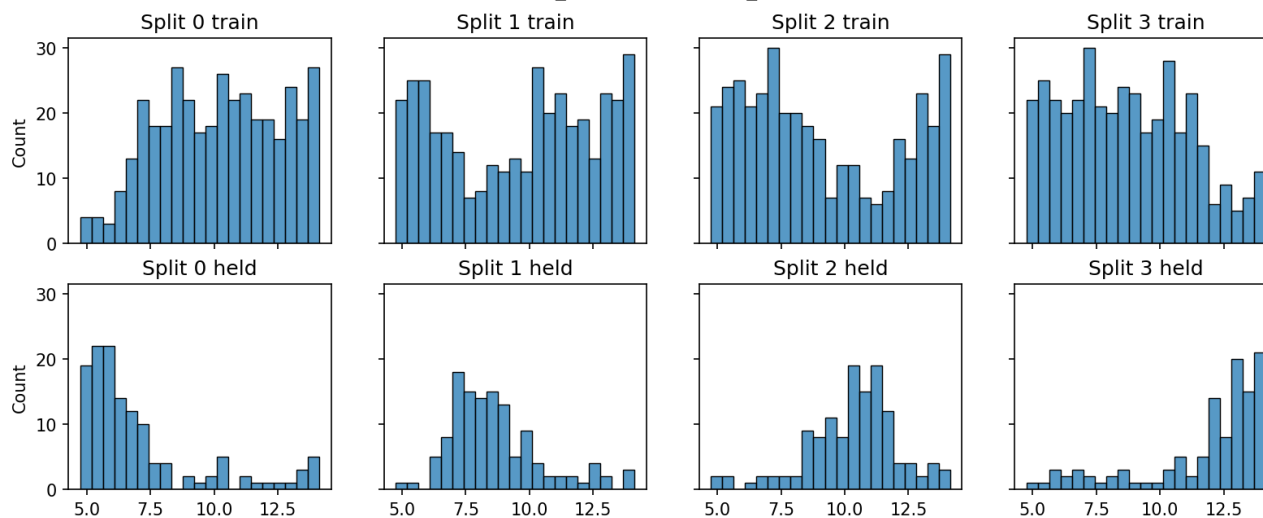


Dataset of Graphs



Split the data based on conditionals

sharpness_scale=10, epsilon_base=0.1



Compute conditionals using Beta distributions with $\psi=10$, and mix in uniform with factor $\epsilon_b=0.1$

