Quiz #9: Recommendation Systems Name: GRADER ID: 1) (4pts) Using item-based CF (N=2) and the Pearson correlation, calculate the rating prediction of item 3 for user 1 using average ratings based on all ratings choose top 2 Peasson Correlations W13, W23 Pulis - [I POLNT] $w_{i,j} = \frac{\sum_{u \in U} (r_{u,i} - \overline{r}_i) (r_{u,j} - \overline{r}_j)}{\sqrt{\sum_{u \in U} (r_{u,i} - \overline{r}_i)^2} \sqrt{\sum_{u \in U} (r_{u,j} - \overline{r}_j)^2}} P_{u,i} = \frac{\sum_{n \in N} r_{u,n} w_{i,n}}{\sum_{n \in N} |w_{i,n}|}$ - W13 XY 11 + W23X Y12 = -0.73×2+(-0.59)Kl 1-0.731+1-0.591 [I POINT] -[2 POINTS] $\begin{cases}
U_{2,1}U_{4}U_{5} \rightarrow \omega_{1/3} = \frac{(3-10/3)_{A}(5-8/3)+(5-\frac{10}{3})(1-8/3)}{\sqrt{(3-10/3)^{2}+(5-10/3)^{2}} \times \sqrt{(5-8/3)^{2}+(1-8/3)^{2}}} \approx -0.43 \\
\frac{1}{\sqrt{(3-10/3)^{2}+(5-10/3)^{2}} \times \sqrt{(5-8/3)^{2}+(1-8/3)^{2}}}{\sqrt{(4-8/3)^{2}+(3-8/3)^{2}} \times \sqrt{(2-8/3)^{2}+(1-8/3)^{2}}} \approx -0.59 \\
\frac{1}{\sqrt{(4-8/3)^{2}+(3-8/3)^{2}} \times \sqrt{(2-8/3)^{2}+(1-8/3)^{2}}}{\sqrt{(4-8/3)^{2}+(3-8/3)^{2}} \times \sqrt{(2-8/3)^{2}+(1-8/3)^{2}}} \approx -0.99 \\
\frac{1}{\sqrt{5-8/3}} \times \frac{(5-8/3) \times (2-8/3) + (2-8/3) \times (3-8/3)}{\sqrt{5-8/3}^{2}+(2-8/3) \times (3-8/3)}} \approx -0.99 \\
\frac{1}{\sqrt{5-8/3}} \times \frac{(5-8/3) \times (2-8/3) \times (3-8/3)}{\sqrt{5-8/3}} \times \frac{(5-8/3) \times (3-8$ ≈ -1.55 2) (2pts) Briefly explain the difference between feature augmentation and meta-level feature augementation - generates a new feature for each item by using the recommendation togic of the contributing domain. eg: tootlent -based model over the training data of then using that model to generate ratings for unvated items.

Meta-level - model learned by one tecommender as input for another, escample-meta-level - model learned by one techniques to bill models of user prof in a content-rustaur rant secommender that used raise bayes techniques to bill models of user prof in a content-rustaur rant secommender that used raise bayes techniques to based way.

3) (4 pts) For node B, use the Girvan-Newman algorithm to calculate the hetweeness of each edge (do this formal algorithm to calculate the hetweeness of each edge (do this for node B ONLY). You need to show the steps of your [2 POINTS]1) Perform a BFS of the graph
Starting node B

a) Label each node by number of
shortest paths that each from calculation. the loot. 3) Each node other than the not gets 4) Fuch node that not the leafgote credit = 1 + sum of credit of
edges. POINT 7 [1 POINT 5) credit is shared for node with more than I parent.