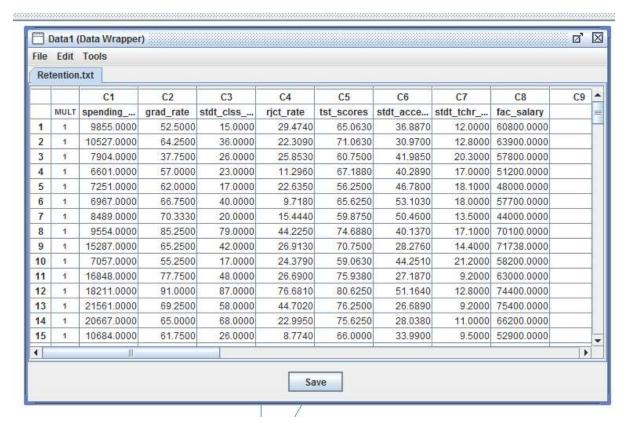
Assignment 7

DATA ANALYTICS – CAUSAL DISCOVERY
LIJU ROBIN GEORGE & JAYASHANKAR MALEPATI

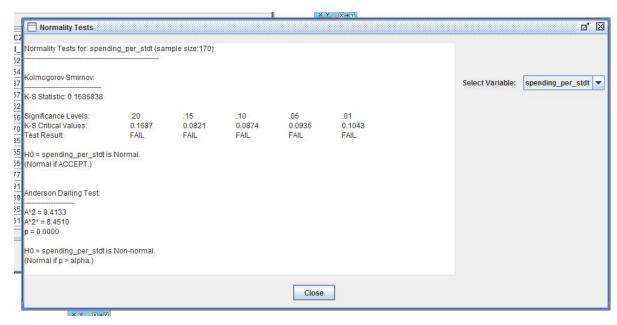
Report:

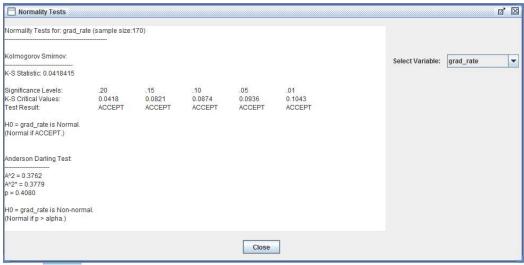
We performed the analysis using tetrad 5.3 as it seemed quite stable and bug free. We did the analysis step by step, as was done in the paper for the 1992 data. At each stage, we compare what we found with what was mentioned in the previous paper.

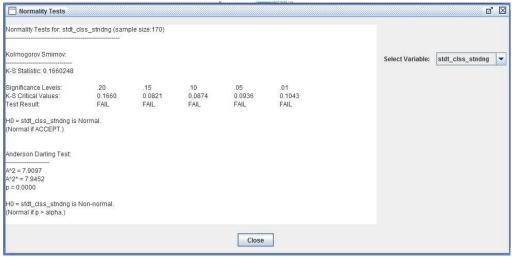
We started off with reading the data in:

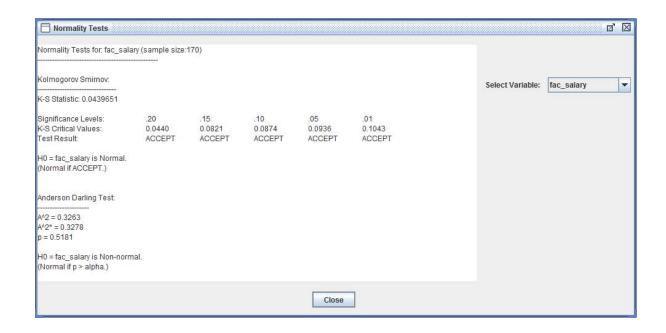


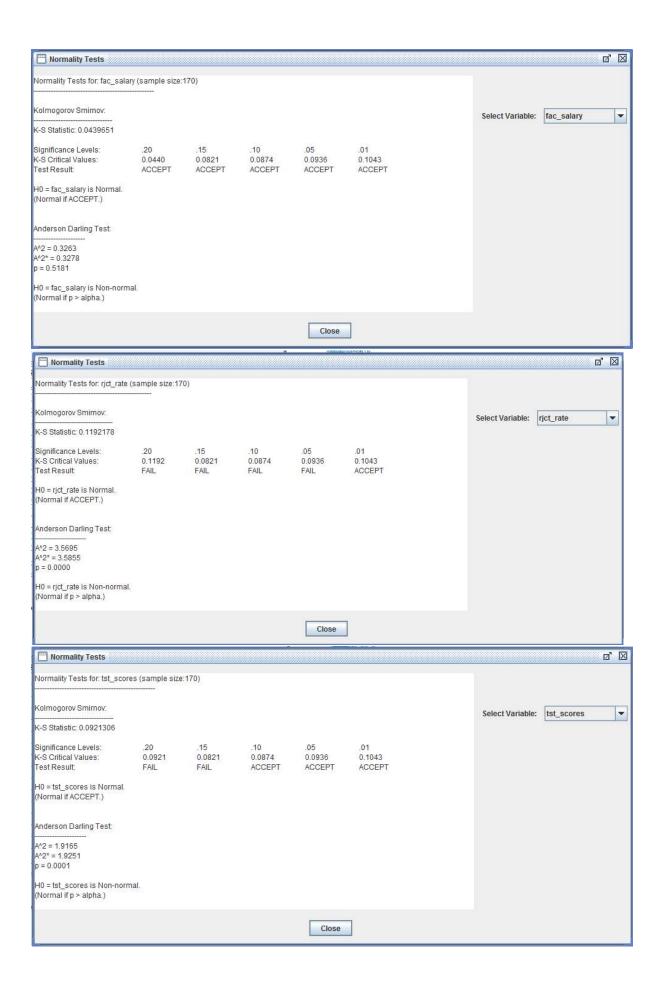
The next step we checked whether the variables of concern were normalized as was mentioned in the paper. We got the following set of results. The normality tests showed that a few variables were not normalize at all, 3 of them specifically spending_pre_stdt, stdt_class_standing, rjct_rate.

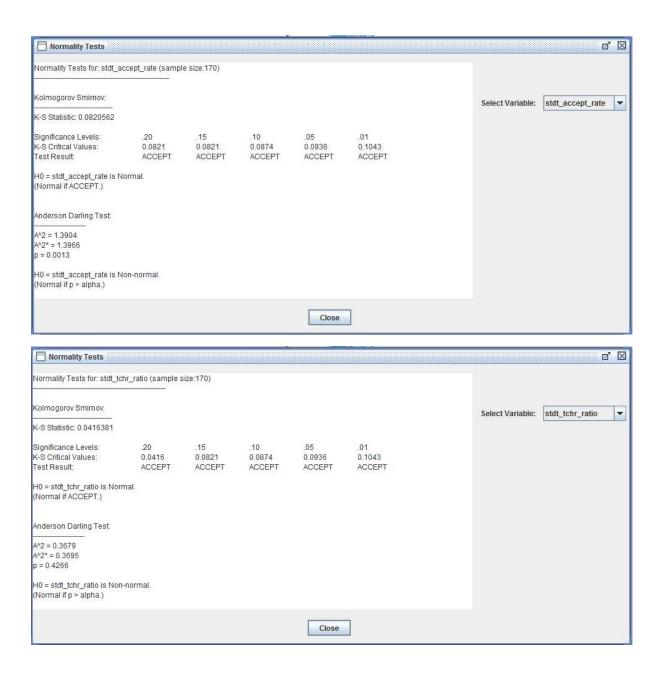




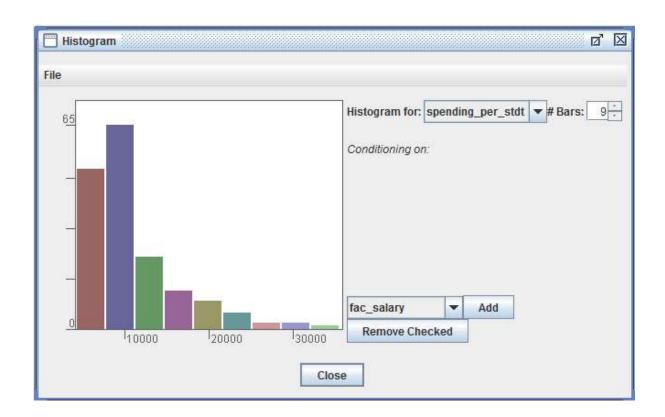




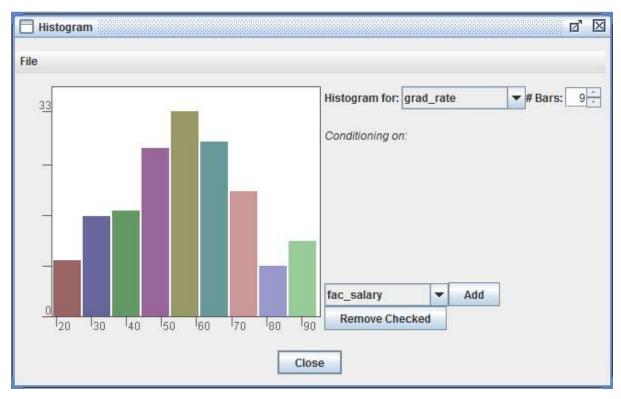




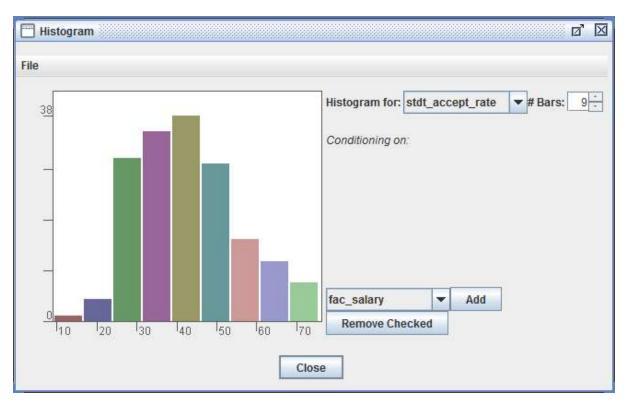
The following histograms also prove this fact. We increased the bin sizes also to see if there were any variations. But did not find much.

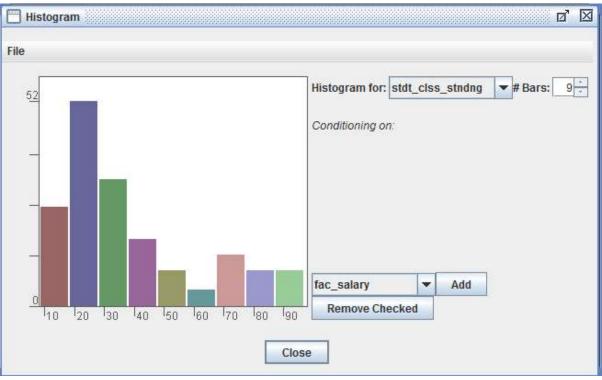


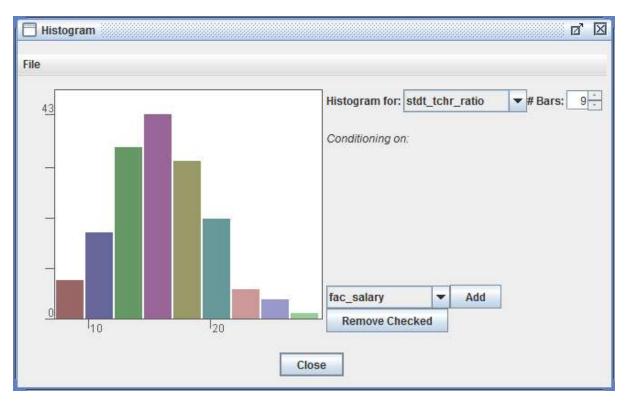














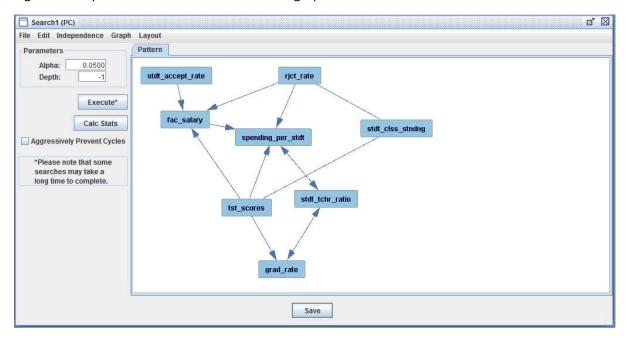
PC Algorithm:

The next step was to run the PC algorithm search for the data that we have. The analysis was done in two stage: Without any pre-knowledge and with pre-knowledge. We tested for various significance levels as shown below.

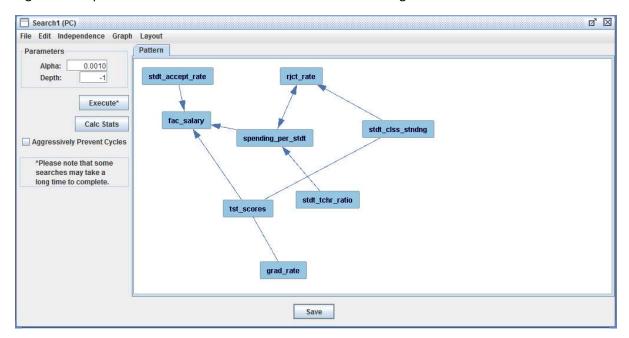
Without Pre-Knowledge:

Without any pre-knowledge, we just put the data into out PC search algorithm and it gave the following causal graph. This was totally random. The results did not show any specific valuable information. Majority of nodes were hitting on spending_stdt_student. We checked this for various significance levels as shown below.

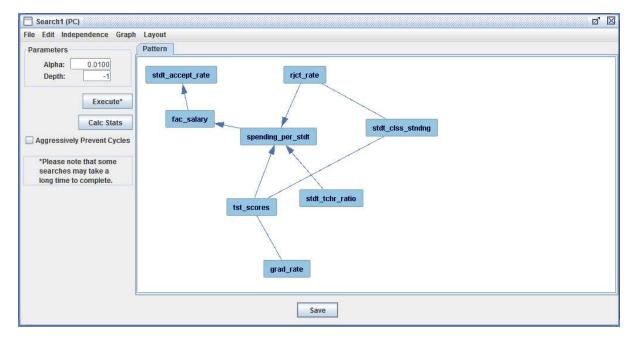
Significance: p = 0.05: Gives a more connected graph.



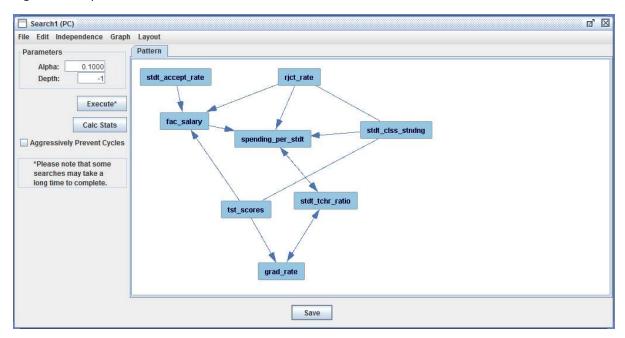
Significance: p= 0.001: Connections are less and some have changed orientations



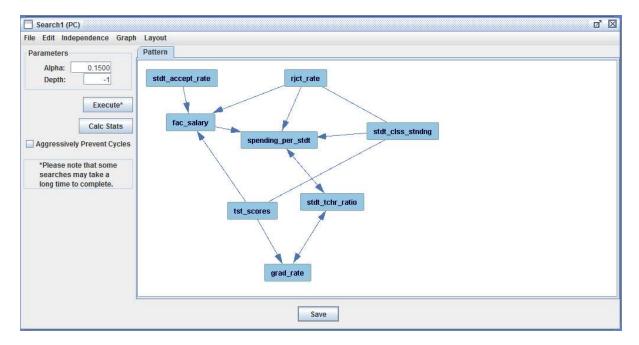
Significance: p= 0.01: Lesser connections and directions



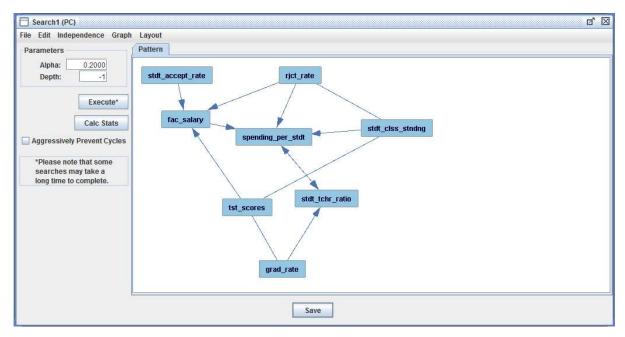
Significance: p= 0.1: Gave more connections but close to 0.05



Significance: p= 0.15

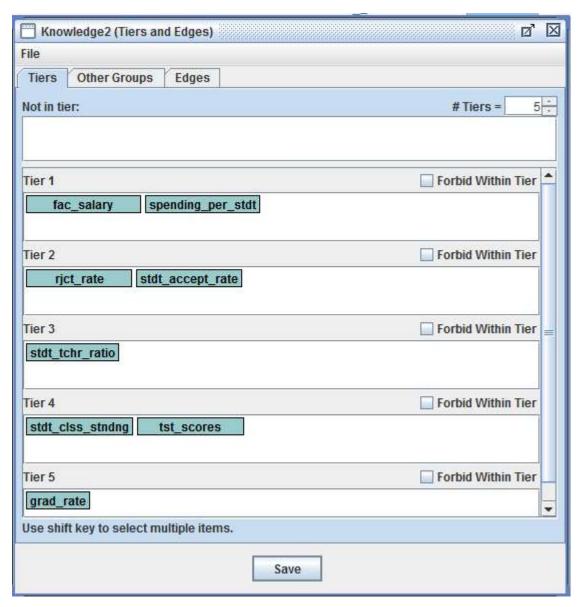


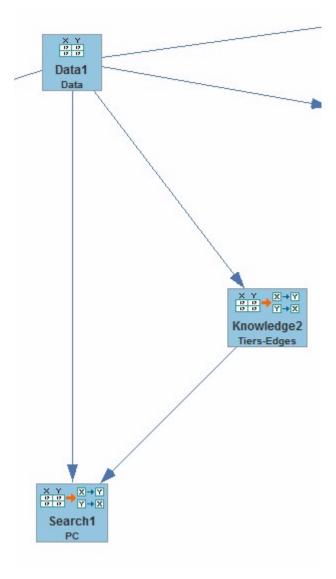
Significance: p= 0.2



With Pre-knowledge: Temporal Order:

Please note we have seen a relation between student_teacher_ratio and graduate rate and decided to put that into a tier just after the student acceptance and reject rates and before the testscores and class standings, because we feel that the student_teacher_ratio would be determined once the number of students joining are decided.



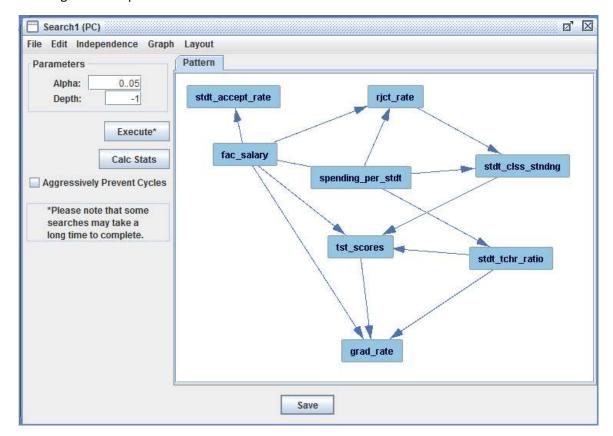


Interestingly we saw that now we have a graph which looked more like the one predicted in the paper. Most variables pass through tst scores and determine the graduation rate.

However, there seems to be a direct relation between std_teacher_ratio and graduation rate as discussed above. This was not seen in the paper for the 1992 data. This relation makes sense too, the better the student_teacher_ratio the more attention a student gets, hence can influence better understanding in class and a higher graduation rate.

We also noted that there is a strange relation shown between fac_salary and rjct_rate. In the analysis below that, we tried removing the edge and seeing the graph. This brought about an interesting change, the connection between std_teacher_ratio and tst_scores were removed, but there is an addition of an edge between spending_per_stdt and tst_scores, which might signal that, fac_salary plays a role in the std_teacher_ratio, spending_per_stdt and hence the tst_scores which again affects the graduation rate.

With Significance: p = 0.05



With Significance: p = 0.05 and edge between fac_salary and rjct_rate removed.

