(c) ARM

(i) two main tasks in ARM

(ii) two key factor: FP-Growth faster than

(iii) metrics

Aprior;

$$(5,0,7) \qquad (0,8,4)$$

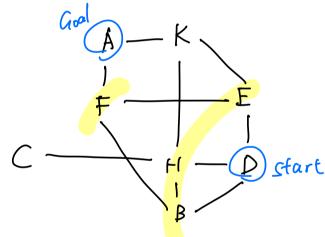
$$(5,7,0) \qquad (5,3,4) \qquad (4,8,0)$$

$$5 \qquad 2 \qquad 4 \qquad (5,5,2) \qquad (0,7,5) \qquad (0,3,9) \qquad (4,0,8)$$

$$1 \qquad 3 \qquad 3 \qquad 1$$

$$(2.8.2)$$
 $(5.2.5)$ $(3.0.9)$ $(0.4.8)$
 $4 \mid 2 \mid 5 \mid (3.8.1)$ $(5.4.3)$
 $(2.0.10)$ $(0.2.10)$ $(3.8.1)$ $(5.4.3)$
 $(5.6.1)$
 $2 \mid (0.6.6)$ $\sqrt{achieve}$ goal

(b) Odepth -first search



Zndex Open Closed.

D []

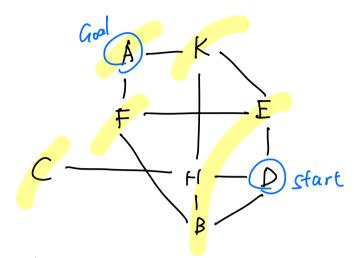
| BEH D

2 FEH BD

3 AEH FBD

DFS use. 4 Steps to find node A

3 Breath First Search



Index	Open	Close d
0	D	[]
	BEH	D
2 .	EHF	BD
3	HFK	EBD
4	FKC	HEBD
5	KCA	FHEBD
6	CA	KFHEBD
7	A	CKFHEBD

BFG use 8 Steps to find node A

- 3 Conclusion: In this graph, the DFS path to A is quite direct, so DFS ends up examining fewer nodes than BFS before finding the goal
 - (c) (i) ① Frequent itemsets Generation:

 Find all the itemsets that satisfy the minsup threshold. These itemsets are called frequent itemsets
- 3 Rule Generation
- Extract all the high-confidence rules from the frequent itemsets. These rules are called Strong rules.
 - (ii) OFP- Growth uses a highly compressed representation of the dataset called on FP-tree which avoid storing many duplicates
 - 2 It avoil the repeated, full-database scans

required by Apriori, Apriori typically scansonce por can didate—itemset size, because it recursively mines conditional FP trees instead.

(iii) O support: frequency of rule's itemset
in the data

② confidence: conditional probability

confidence(x > Y) = support(x b Y)

support(x)

(3) If $f : \text{Lift}(X \rightarrow Y) = \frac{\text{confidence}(X \rightarrow Y)}{\text{Suppore}(Y)}$