191 Name Noman khan Skat ne: 820 102 131 Question 4.1: Solution: Given likelihoods: neg 0.09 0.16 always 0.07 0.06 like 0.29 0.06 foreign 0.04 0.15 films 0.08 0.11 Prior Probabilities (equal): P(pos) = P(neg) = 0.5 P(pos) 'I always like foreign films')=
P(pos) x P('I') pos) x P(salways') pos)
x P('like') pos) x P('foreign') pos) x
P('films') pos) =) P(pos 1 'I always like foreign films') = 0.5 x 0.07 x 0.29 x 0.04 x 0.08 =) P(pos 1° I always lete foreign filmi') = 0.00053248

P(neg l'I always like foreign films')=0.5x 0.96 x 0.06 x 0.06 x 0.15 x 0.11 =) P(neg l'I always like foreign films')= 0.0000297 P(pos 1'I always like foreign films')

> P(neg 1'I always like foreign

films) : Naive Bayes would Classify the Sentence I always like foreign films as positive. Duestion 4.2: Solution: Step 1: Computing Prior Probabilities: P(Comedy) = 2 = 0.4 P(action) = 3 = 0.6Comedy Class: words fun Count

Couple love Total words in Cornedy cluss: 9 Vocabulary (unique words) Size: Action class: Count Words fast furious Shoot love Total words in action class: 11 Vocabulary: { fun, Couple, love, fly, fast, (words in fusious, Shoot 3 both classes Itotal words) Vocabulary Size: 7 Likeli hoods with Add- 1 Smoothing: P(word | Class) = Count (word in Class) total words + Vocabrelay

Cornedy Class likelihoods: P (fast | Cornedy) = 1+1 = 2 = 0.125 P(Carple | Canedy) = 2+1 = 3 = 0.1875 9+7 16 P(Shoot | Comedy) = 0+1 = 1 = 0.0625 P(fly | Comedy) = 1+1 = 2 = 0.125 Action Class liteli hoods: P(fast | action) = 2+1 = 3 = 0.166 11+7 = 48P(Couple |action) = 0+1 = 1 = 0.055 P (Shoot |action) = 4+11 = 15 = 0.833 11+7 18

Pg 5 P(fly | action) = 1+1 = 2 = 0.11 11+7 18 Compute Posteriar probabilitées: P(Cornedy ID) = P(Cornedy) x P(fest Kornedy) x P(Couple | Cornedy) x document PCShoot (Cornedy) x Ply (Comedy)  $= 0.4 \times 0.125 \times 0.1875 \times 0.0625$ x 0 125 = 0.000073P(action | D) = 0.6 x 0. 166 x 0. 055x 0.833 = 0.00456 " P (action 10) > P (Comedy 10) .. The most likely Class for document D is action.

Question 4.3: Solution: Given document data: doc 'good' 'poor' 'great' class ds 0 Vocabulary = { good, poor, great } Votabulary Size = 3 Prior Probabilities: P(pos) = Npos (no. of + docs).

N. (Total docs)  $=\frac{2}{R}=\boxed{0.4}$ P(neg) = 3 = 0.6

( for multinomial NB) Litelihoods with Add-1 Smoothing: P(word | Class) = no. of times word + 1 total words + Vocat in class Size Ctotal words in all close W/o repitition P(cgood'|pos) = 3+1 = 4 = 0.33 9+3 12 P('poon' | pos) = 1+1 = 2 = 0.166 9+3 12  $P(\text{great}^2 | \text{pos}) = 5+1 = 6 = [0.5]$   $9+3 \quad 12$  $P(cgood' | neg) = \frac{a+1}{14+3} = \frac{3}{17} = 0.176$ P('paon' | neg) = 10+1 = 11 = (0.647 14+3 17

Classify the Sentence: Using multinomial NB: P(pos | Sentence) = P(pos) x P ( good | pos) x disp words that P( good 1 pos)x are not present P((poor / pos) x in either class P('great' pos) = 0.4 x (0.33) x 0.166 x 0.5 = 0.00361P (neg | Sentence) = 0.6 x (0.176) x 0.647x 0.476 = 0.00211 : P(neg | Sentence) > P(pos | Sentence).
: multinomial NB would Classify.
the Sentence as Negative. Using Binarized NB: roultible Prob for Samepos) x P(cpoor' | pos) x
words are ignored in P(cgreat' | pos)

- 0.4 x 0.5 x 0.5 x 0.75 = 0.075 P(neg | Sentence) = P(neg) x
P('good' | neg) x
P('poor' | neg) x P('great'  $= 0.6 \times 0.6 \times 0.8 \times 0.4$  = 0.1152: P(neg | Sentence) > P(pos | Sentence)
: Binary NB would also classify
the Sentence as Negative. Conclusion: Both the models agrel on that the given Sentence is a negative Sentence.