- 1) Naïve Bayes for Text Classification
- → Classes: flower, vegetable, fruit

$$P(c) = Nc/N$$

P(w|c) = Count(w,c)/Count(c) + kV

Class probabilities:

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P(flower) = 3/8

P(vegetable) = 2/8

P(fruit) = 3/8
```

For Document 1:

→ Conditional probabilities: D1: rose, lily, apple, carrot

Therefore, since P(fruit|D1) is the highest for document D1, the most likely class for D1 is **fruit.**

For Document 2:

→ Conditional probabilities: D2: pea, pea, lotus, grape

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P(\text{pea |flower}) = (1+0.1)/[13+(0.1*14)] = 1.1/14.4 \\ P(\text{lotus |flower}) = (0+0.1)/[13+(0.1*14)] = 0.1/14.4 \\ P(\text{grape |flower}) = (0+0.1)/[13+(0.1*14)] = 0.1/14.4 \\ P(\text{pea |vegetable}) = (2+0.1)/[8+(0.1*14)] = 2.1/9.4 \\ P(\text{lotus |vegetable}) = (1+0.1)/[8+(0.1*14)] = 1.1/9.4 \\ P(\text{grape |vegetable}) = (0+0.1)/[8+(0.1*14)] = 0.1/9.4 \\ P(\text{pea |fruit}) = (0+0.1)/[14+(0.1*14)] = 0.1/15.4 \\ P(\text{lotus |fruit}) = (1+0.1)/[14+(0.1*14)] = 1.1/15.4 \\ P(\text{grape |fruit}) = (2+0.1)/[14+(0.1*14)] = 2.1/15.4 \\ P(\text{flower}|D1) = 3/8*1.1/14.4*1.1/14.4*0.1/14.4*0.1/14.4*0.1/14.4 = 0.00000011 \\ P(\text{vegetable}|D1) = 2/8*2.1/9.4*2.1/9.4*1.1/9.4*0.1/9.4 = 0.00000015 \\ P(\text{fruit}|D1) = 3/8*0.1/15.4*0.1/15.4*1.1/15.4*2.1/15.4 = 0.000000015 \\ P(\text{fruit}|D1) = 3/8*0.1/15.4*0.1/15.4*1.1/15.4*2.1/15.4*2.1/15.4 = 0.000000015 \\ P(\text{fruit}|D1) = 3/8*0.1/15.4*0.1/15.4*1.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*2.1/15.4*
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Therefore, since P(vegetable | D2) is the highest for document D2, the most likely class for D2 is **vegetable**.