Exercises 2

Relationships between events illustrated.

Two screening test are available for identifying loan defaulters. These tests are used to evaluate requests for new loans.

Test A – Credit rating result: good (G) or poor (P)

Test B - Employment Status result: satisfactory (S) or not satisfactory (NS)

Statistical analysis has provided the following estimates of the test performances produced the following tables:

Sample size = 10158 loans

Default = D, non default = \overline{D}

| | Test A | |
|---|--------|----------------|
| | D | \overline{D} |
| Р | 135 | 1008 |
| | | |
| G | 255 | 8760 |

| | Test B | |
|----|--------|----------------|
| | D | \overline{D} |
| NS | 120 | 444 |
| | | |
| S | 270 | 9324 |

We shall estimate the probabilities from this large sample.

Dividing the cells of A by the sample size we get the following probabilities:

| | Test A | |
|---|----------|----------------|
| | D | \overline{D} |
| Р | 0.01329 | 0.099232 |
| | | |
| G | 0.025103 | 0.862374 |

What we have in the table is that:

$$P((A = P) \cap D) = 0.01329$$
 etc

P(D) = 0.01329 + 0.025103 = 0.038393 or nearly 4% of loans default.

Accuracy of the test at picking up potential defaulters is described by:

$$P(D \mid P) = \frac{P(D \cap P)}{P(P)} = \frac{0.01329}{0.01329 + 0.099232} = 0.11811$$

Thus 12% of people with poor credit rating default, in 88% of these cases the loan is OK.

Suppose the evaluating software applies this criterion rejecting those with the result P.

$$P(D \mid G) = \frac{P(D \cap G)}{P(G)} = \frac{0.025103}{0.025103 + 0.862374} = 0.028286$$

A reduction in the bad credit rate from 3.8% to 2.8%. However customers will be lost:

$$P(P) = 0.01329 + 0.099232 = 0.112522$$
, 11.3% of customers are rejected.

$$P(\overline{D} \mid P) = \frac{0.099232}{0.112522} = 0.88199$$
, 88.2% of these would have been OK.

Suppose an OK loan nets the company \$2000, the loss on a defaulted loan is \$10000.

If they don't use the criterion they will make on (large) N loan applications:

$$N * (0.038393* (-10000) + 0.961607*2000) = 1539.279*N$$

If they do then

So they should leave well alone or get better info they are rejecting too many.

Exercises:

As exercise evaluate Test B in the same way.

If you could only use one test which would you use?

Using both the tests.

We now have 3 events (P,G), (S,NS) and (D, \overline{D})

The two tables above do not (in general) contain enough information for the 3 event model. What is the relationship between the tests?

$$P(P \cap S) = ?$$

How do the tests jointly impact on P(D)?

The full model involves 7 parameters; suppose the disaggregated table is :

| | Defaulting | | |
|--------|------------|-----|----------------|
| Test A | Test B | D | \overline{D} |
| | NS | 108 | 432 |
| Р | | | |
| | S | 27 | 576 |
| | | | |
| G | NS | 12 | 12 |
| 9 | S | 243 | 8748 |

Note: the data in this table is the disaggregated version of the data in the tables above.

So that P(D) = 0.038393 is the same as before.

We can get the probabilities as:

$$P(P \cap NS \cap D) = \frac{108}{10158} \quad \text{etc}$$

Exercises:

Compute the probabilities of defaulting and of rejecting a potentially good client with the following strategies:

- 1. Only grant a loan if pass both tests.
- 2. Only reject if fail both tests.

Are the tests independent?