## Handout 12: Contingency Tables

While there are many uses of the G-test, the most common application is in the study of contingency tables. Consider, for example, a multinomial with $k=4$, just as before. However, this time we are going to arrange the data into a two-by-two table, using a slightly different notation for the counts to make it clear that each is associated with a specific row and column. This yields the following, where we have added row sums $r_{j}$ and column sums $c_{j}$, since we will need them in a moment:

| $x_{1,1}$ $x_{1,2}$ <br> $r_{1}$  <br> $x_{2,1}$ $x_{2,2}$ <br> $r_{2}$  <br> $c_{1}$ $c_{2}$ | $n$ |
| :---: | :---: | :---: |

We can re-define the multinomial probabilities similarly, where $p_{i, j}$ is the probability of landing in row $i$ and column $j$. A very common type of hypothesis test is to consider the set $\Theta_{0}$ of all tables in which event of being in row $i$ is independent of the event of being in column $j$, for all combinations of $i$ and $j$.

The maximum likelihood estimator is unchanged in this case; it is still the raw counts divided by the sample size. The numerator of the $G$ test, however, is different. In order to be in $\Theta_{0}$, we need to have that $p_{i, j}$ is equal to the probability of being in row $i$ times the probability of being in column $j$. It should not be surprising to know then that in order to maximize the log-likelihood under $H_{0}$, we use the following probabilities and implied expected counts:

$$
\tilde{p}_{i, j}=\left(\frac{r_{i}}{n}\right) \times\left(\frac{c_{j}}{n}\right) \Rightarrow e_{i, j}=\left(\frac{r_{i} \times c_{j}}{n}\right) .
$$

In other words, the proportion of data that were in row $i$ times the proportion of data that were in column $j$. From here, we use the same formula as we have on the other page by replacing the sum of $j$ with a double sum over both $i$ and $j$.

We can extend this same approach to the case where we have $R$ rows and $C$ columns. What, in general, will be the degrees of freedom for $G$ ? We have $C R-1$ dimensions in $\Theta$ (any set of probabilities, with the one restriction that the sum to 1 ) and $(C-1)+(R-1)$ in $\Theta_{0}$ (any set of valid column probabilities and row probabilities, each having to sum to 1 ). This difference factors as:

$$
(C R-1)-(C-1)-(R-1)=(C-1) \cdot(R-1) .
$$

So, in the common two-by-two table case, we have only a single degree of freedom. This will grow larger for tables with more rows and/or columns.

