Today

- 9.4: Generalized Likelihood Ratio Test
- 9.3: Duality Between Tests and Confidence Intervals

9.4 Generalized Likelihood Ratio Tests
The Neyman-Pearson likelihood ratio test is most powerful for simple vs. simple.
Here, we generalize to composite hypotheses. The generalized LRT is not necessarily optimal, but works well for situations where no optimal test exists.
Observe $X_1, \ldots, X_n$ from $f(x|\theta)$.
Test $H_0$: $\theta \in \omega_0$ $\omega_0 \subset \Omega$
vs. $H_A$: $\theta \in \omega_1$ ($\omega_0 \cup \omega_1 = \Omega$).
Use the generalized LR statistic.

$$\Lambda = \frac{\max_{\theta \in \omega_0} (\text{lik}(\theta))}{\max_{\theta \in \Omega} (\text{lik}(\theta))} = \frac{\max_{\theta \in \omega_0} \prod_{i=1}^{n} f(x_i|\theta)}{\max_{\theta \in \Omega} \prod_{i=1}^{n} f(x_i|\theta)}$$

Reject $H_0$ if $\Lambda \leq \lambda_0$, where $\lambda_0$ is
$P(\text{rej. } H_0|\theta \in \omega_0) = \alpha$.
(Note: If $H_0$ holds, $\Lambda = 1$. If $H_A$ holds, $\Lambda < 1$, small).
Use this to construct the test, i.e. find rejection regions in terms of simple statistics (similar to N-P lemma).

Example: Consider a random sample from a normal distribution with unknown mean and unknown variance.
Theorem A pg.341: Under smoothness conditions on the pdf, the null distribution of \(-2\ln \Lambda\) has an approximate chi-square distribution with d.f.\(=\dim \Omega - \dim \omega_0\) for large \(n\).

\(N(\mu, \sigma^2)\), both unknown
\(\Rightarrow\) df = 2 - 1 = 1.

9.3 The Duality of Confidence Intervals and Hypothesis Tests

There is a duality between confidence intervals and hypothesis tests. A confidence interval is found by “inverting” a two-sided test (and vice-versa).

Theorem A, pg. 338: Suppose there is a test of level \(\alpha\) for \(H_0: \theta = \theta_0\), and let \(A(\theta_0)\)=acceptance region.

Then the set \(C=\{\theta: X \in A(\theta)\}\) is a 100(1 - \(\alpha\))% confidence region for \(\theta\).
Theorem B, pg. 338: Let $C(X)$ be a $100(1 - \alpha)\%$ confidence region for $\theta_0$.

Then $A(\theta_0) = \{X: \theta_0 \in C(X)\}$ is an acceptance region for a test of level $\alpha$ for $H_0: \theta = \theta_0$.

Example cont.: Again, consider the random sample from a normal distribution with unknown mean and unknown variance.