

# Likelihood Ratio, Wald, and Lagrange Multiplier (Score) Tests

Soccer Goals in European Premier  
Leagues - 2004

# Statistical Testing Principles

- Goal: Test a Hypothesis concerning parameter value(s) in a larger population (or nature), based on observed sample data
- Data – Identified with respect to a (possibly hypothesized) probability distribution that is indexed by one or more unknown parameters
- Notation:

Data:  $y_1, \dots, y_n$

Parameter(s):  $\theta_1, \dots, \theta_k$

Joint Density Function:  $f(y_1, \dots, y_n | \theta_1, \dots, \theta_k)$

## Example – English League – Total Goals/Match

- Suppose we wish to test whether the mean number of goals (in a hypothetically infinite population) of games is equal to 3. Note: all games of equal length (no overtime in regular season games)
- Data:  $Y$ =Total # of goals in a randomly selected game
- Distribution: Assume Poisson with parameter  $\theta$
- Null Hypothesis:  $H_0: \theta = 3$
- Alternative Hypothesis:  $H_A: \theta \neq 3$
- Joint Probability Density Function:

$$f(y_1, \dots, y_n | \theta) = \prod_{i=1}^n \frac{e^{-\theta} \theta^{y_i}}{y_i!} = \frac{e^{-n\theta} \theta^{\sum_{i=1}^n y_i}}{\prod_{i=1}^n y_i!} \quad y_i = 0, 1, 2, \dots$$

# Likelihood Function

- Another term for joint probability density/mass function. Common Notation:  $L(\theta)$  or  $L(\theta, y)$  or  $L(\theta|y)$
- Considered as a function of both the (observed) data and the (unknown) parameter values
- Used in estimation and testing parameter value(s)
- Goal is to choose parameter value(s) that maximize likelihood function given the observed data.
- Typically work with the log of the likelihood, as it is often easier to differentiate to solve for maximum likelihood (ML) estimators for many families of probability distributions

# ML Estimation of Poisson Mean

$$L(\theta, y) = \frac{e^{-n\theta} \theta^{\sum_{i=1}^n y_i}}{\prod_{i=1}^n y_i!}$$

$$l = \ln(L(\theta, y)) = -n\theta + \left( \sum_{i=1}^n y_i \right) \ln(\theta) - \ln \left( \prod_{i=1}^n y_i! \right)$$

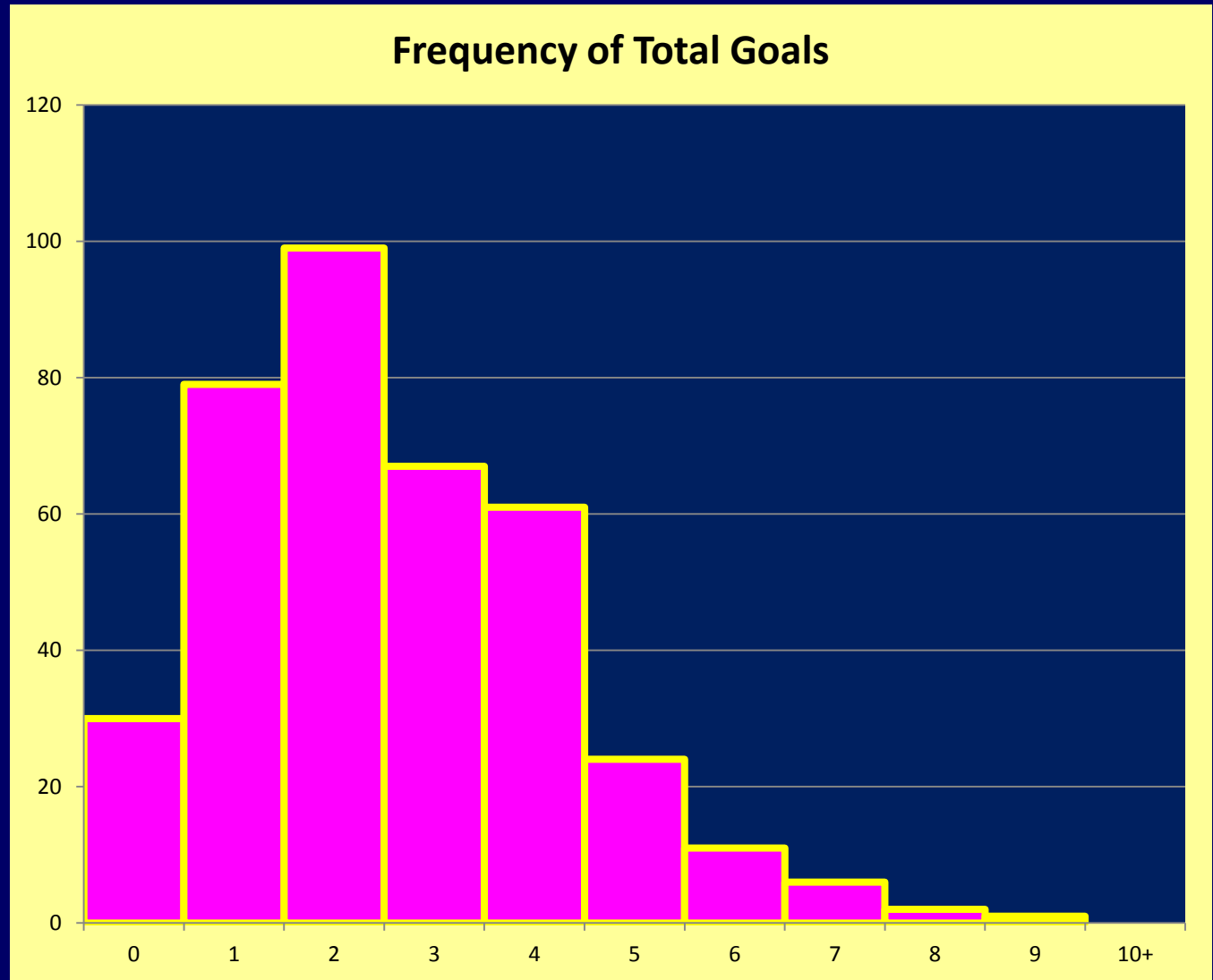
Taking derivative (wrt  $\theta$ ) and setting to zero for maximum:

$$\frac{dl}{d\theta} = -n + \frac{\sum_{i=1}^n y_i}{\theta} \stackrel{\text{set}}{=} 0 = 0 \quad \Rightarrow \quad -n + \frac{\sum_{i=1}^n y_i}{\hat{\theta}} = 0 \quad \Rightarrow \quad \hat{\theta} = \frac{\sum_{i=1}^n y_i}{n} = \bar{y}$$

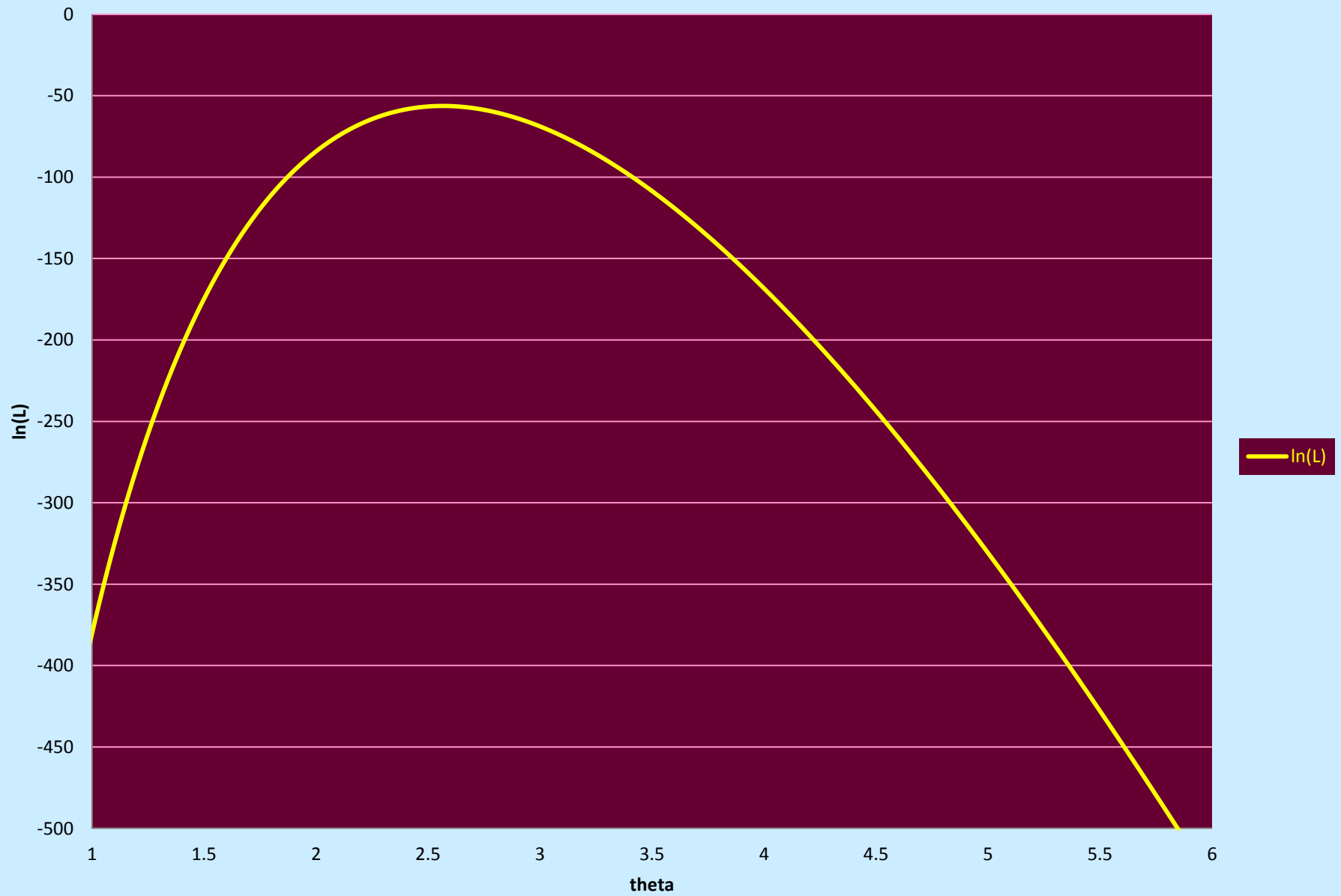
# Total Goals Data

Goals	Frequency
0	30
1	79
2	99
3	67
4	61
5	24
6	11
7	6
8	2
9	1
10+	0
<b>Total</b>	<b>380</b>

$$\hat{\theta} = \frac{\sum_{i=1}^{380} y_i}{380} = \frac{975}{380} = 2.57$$



**ln(L) versus theta (Ignoring constant term)**



# Likelihood Ratio Test

- Identify the parameter space:  $\Omega = \{\theta: \theta > 0\}$
- Identify the parameter space under  $H_0$ :  $\Omega_0 = \{\theta: \theta = \theta_0\}$
- Evaluate the maximum log-Likelihood
- Evaluate the log-Likelihood under  $H_0$
- Any terms not involving parameter can be ignored
- Take -2 times difference ( $H_0$  – maximum)
- Under null hypothesis (and large samples), statistic is approximately chi-square with 1 degree of freedom (number of constraints under  $H_0$ )

$$X_{LR}^2 = -2 \left[ \ln \left( L(\theta_0, y) \right) - \ln \left( L(\hat{\theta}, y) \right) \right]$$



# Soccer Goals Example

$$\ln(L(\theta, y)) = -380\theta + \left(\sum_{i=1}^{380} y_i\right) \ln(\theta) - \ln\left(\prod_{i=1}^{380} y_i!\right)$$

Under  $H_0 : \theta = 3$  (Ignoring  $\ln\left(\prod_{i=1}^{380} y_i!\right)$ ):

$$\ln(L(\theta = 3, y)) = -380(3) + 975(\ln(3)) = -1140 + 1071.15 = -68.85$$

Maximum Value @  $\hat{\theta} = 2.57$ :

$$\ln\left(L\left(\hat{\theta}, y\right)\right) = -380(2.57) + 975(\ln(2.57)) = -976.6 + 920.31 = -56.29$$

Test Statistic:

$$X_{LR}^2 = -2 \left[ \ln(L(\theta = 3, y)) - \ln\left(L\left(\hat{\theta}, y\right)\right) \right] = -2(-68.85 - (-56.29)) = 25.12 > \chi_{.05,1}^2 = 3.84$$

We have strong evidence to conclude the “true” mean total number of goals is below 3.

# Wald Test - I

- By Central Limit Theorem arguments, many estimators have sampling distributions that are approximately normal in large samples
- Then, if we have an estimate of the variance of the estimator, we can obtain a chi-square statistic by taking the square of the distance between the ML estimate and the value under  $H_0$  divided by the estimated variance
- The estimated variance can be obtained from the second derivative of the log-Likelihood

# Wald Test - II

$$V\left(\hat{\theta}\right) = \frac{1}{n} I^{-1}(\theta) \quad \text{where: } I(\theta) = -\frac{1}{n} E\left[\frac{\partial^2 \ln(L)}{\partial \theta^2}\right]$$

$$\text{Wald Chi-Square Statistic: } X_w^2 = \frac{\left(\hat{\theta} - \theta_0\right)^2}{\hat{V}\left(\hat{\theta}\right)} = nI\left(\hat{\theta}\right)\left(\hat{\theta} - \theta_0\right)^2$$

$$\text{Poisson Model: } \ln(L(\theta, y)) = -n\theta + \left(\sum_{i=1}^n y_i\right) \ln(\theta) - \ln\left(\prod_{i=1}^n y_i!\right)$$

$$\Rightarrow \frac{\partial \ln(L(\theta, y))}{\partial \theta} = -n + \frac{\sum_{i=1}^n y_i}{\theta} \quad \Rightarrow \quad \frac{\partial^2 \ln(L(\theta, y))}{\partial \theta^2} = -\frac{\sum_{i=1}^n y_i}{\theta^2}$$

$$\Rightarrow I(\theta) = -\frac{1}{n} E\left[\frac{\partial^2 \ln(L)}{\partial \theta^2}\right] = -\frac{1}{n} \left(-\frac{n\theta}{\theta^2}\right) = \frac{1}{\theta}$$

$$\Rightarrow X_w^2 = \frac{\left(\hat{\theta} - \theta_0\right)^2}{\hat{V}\left(\hat{\theta}\right)} = nI\left(\hat{\theta}\right)\left(\hat{\theta} - \theta_0\right)^2 = \frac{n\left(\hat{\theta} - \theta_0\right)^2}{\hat{\theta}} = \frac{380(2.57 - 3)^2}{2.57} = 27.34$$

# Lagrange Multiplier (Score) Test

- Obtain the first derivative of the log-Likelihood evaluated at the parameter under  $H_0$  (This is the slope of the log-Likelihood, evaluated at  $\theta_0$  and is called the **score**)
- Multiply the square of the score by the variance of the ML estimate, evaluated at  $\theta_0$ . This is the inverse of the variance of the score.
- Then chi-square test statistic is computed as follows:

$$X_{LM}^2 = \frac{s(\theta_0, y)^2}{nI(\theta_0)} \quad \text{where } s(\theta, y) = \frac{\partial \ln(L(\theta, y))}{\partial \theta}$$

# Soccer Goals Example

$$\ln(L(\theta, y)) = -n\theta + \left(\sum_{i=1}^n y_i\right) \ln(\theta) - \ln\left(\prod_{i=1}^n y_i!\right)$$

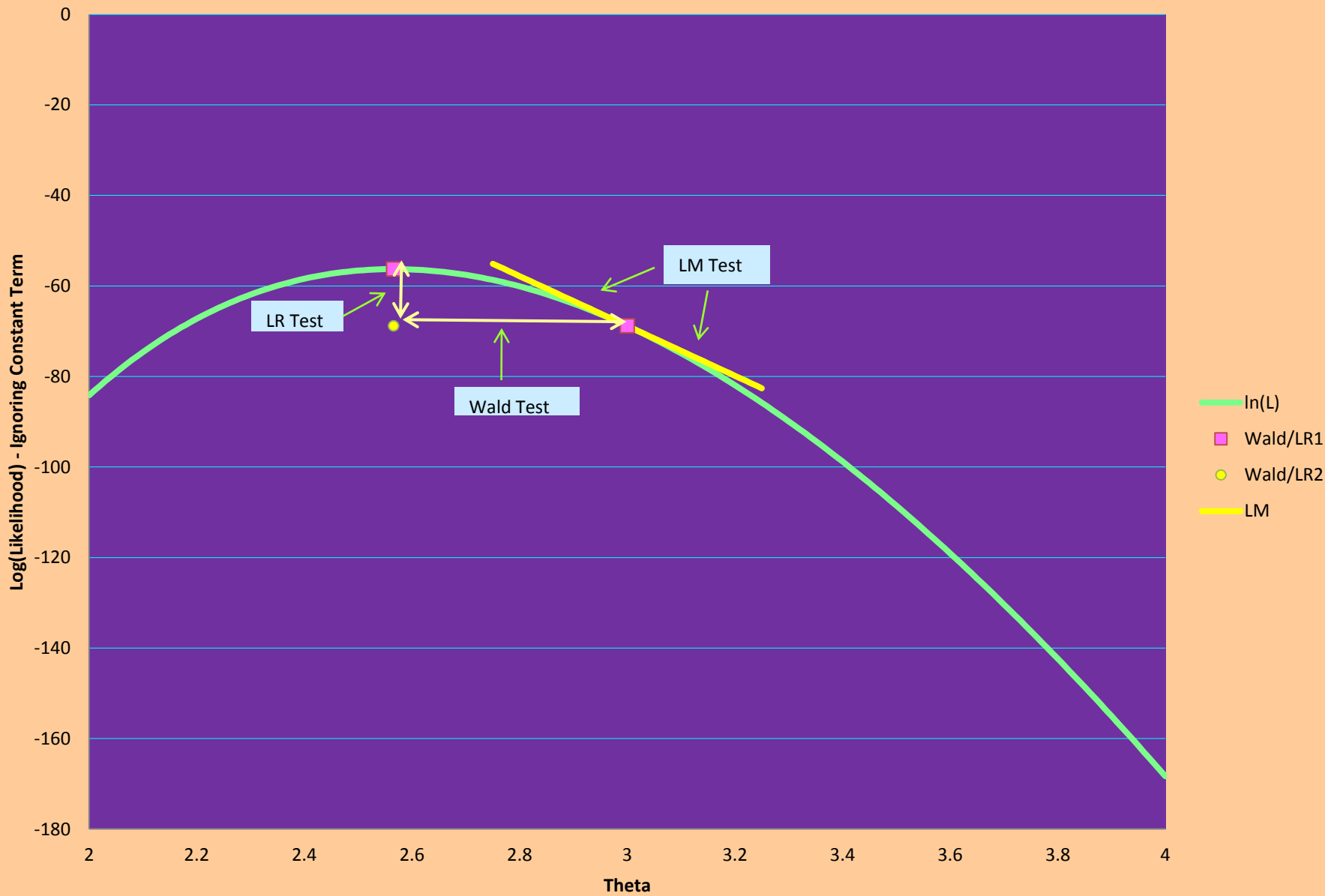
$$\Rightarrow s(\theta, y) = \frac{\partial \ln(L(\theta, y))}{\partial \theta} = -n + \frac{\sum_{i=1}^n y_i}{\theta} \Rightarrow s(\theta_0, y) = -380 + \frac{975}{3} = -55$$

$$I(\theta) = \frac{1}{\theta} \Rightarrow I(\theta_0) = \frac{1}{\theta_0} = \frac{1}{3}$$

$$\Rightarrow X_{LM}^2 = \frac{(s(\theta_0, y))^2}{nI(\theta_0)} = \frac{(-55)^2}{385(1/3)} = 23.57$$

Note that:  $X_W^2 = 27.34 > X_{LR}^2 = 25.12 > X_{LM}^2 = 23.57$

# Log-Likelihood versus Theta (Ignoring Constant Term)



# Generalization to Tests of Multiple Parameters

$$\text{Parameter Vector: } \theta = \begin{bmatrix} \theta_1 \\ \vdots \\ \theta_k \end{bmatrix} \quad H_0 : R\theta = r \quad R = \begin{bmatrix} R_{11} & \cdots & R_{1k} \\ \vdots & \ddots & \vdots \\ R_{g1} & \cdots & R_{gk} \end{bmatrix} \quad r = \begin{bmatrix} r_1 \\ \vdots \\ r_g \end{bmatrix} \quad \text{rank}(R) = g$$

Maximum Likelihood Estimator over entire parameter space:  $\hat{\theta}$

Maximum Likelihood Estimator over constraint under  $H_0$ :  $\tilde{\theta}$

$$\text{Likelihood Ratio Statistic: } X_{LR}^2 = -2 \left[ \ln \left( L(\tilde{\theta}, y) \right) - \ln \left( L(\hat{\theta}, y) \right) \right]$$

$$\text{Wald statistic: } X_W^2 = n_{\bullet} \left( R\hat{\theta} - r \right)^T \left( RI^{-1}(\hat{\theta})R^T \right)^{-1} \left( R\hat{\theta} - r \right)$$

$$\text{Lagrange Multiplier (Score) Statistic: } X_{LM}^2 = \frac{1}{n_{\bullet}} s(\tilde{\theta}, y)^T \left( I(\tilde{\theta}) \right)^{-1} s(\tilde{\theta}, y)$$

$$\text{where: } I_{ij}(\theta) = -\frac{1}{n_{\bullet}} E \left[ \frac{\partial^2}{\partial \theta_i \partial \theta_j} \ln(L(\theta, y)) \right] \quad s_i(\theta, y) = \frac{\partial}{\partial \theta_i} \ln(L(\theta, y))$$

# Soccer Goals Example

- Premier League Games in 2004 for  $k=5$  European Countries:

- England  $n_1 = 380, Y_{1\bullet} = 975$
- France  $n_2 = 380, Y_{2\bullet} = 826$
- Germany  $n_3 = 306, Y_{3\bullet} = 890$
- Italy  $n_4 = 380, Y_{4\bullet} = 960$
- Spain  $n_5 = 380, Y_{5\bullet} = 980$

$$L(\theta, y) = \frac{\exp\left(-\sum_{i=1}^5 n_i \theta_i\right) \prod_{i=1}^5 \theta_i^{y_{i\bullet}}}{\prod_{i=1}^5 \prod_{j=1}^{n_i} y_{ij} !} \quad y_{i\bullet} = \sum_{j=1}^{n_i} y_{ij}$$



# Testing Equality of Mean Goals Among Countries - I

$$H_0 : \theta_1 = \theta_2 = \theta_3 = \theta_4 = \theta_5 \Rightarrow R\theta = r \quad R = \begin{bmatrix} 1 & -1 & 0 & 0 & 0 \\ 1 & 0 & -1 & 0 & 0 \\ 1 & 0 & 0 & -1 & 0 \\ 1 & 0 & 0 & 0 & -1 \end{bmatrix} \quad r = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\ln(L(\theta, y)) = -\sum_{i=1}^5 n_i \theta_i + \sum_{i=1}^5 (y_{i\cdot}) \ln(\theta_i) - \ln\left(\prod_{i=1}^5 \prod_{j=1}^{n_i} y_{ij}!\right)$$

$$\text{Under } H_0 : \ln(L(\theta, y)) = -\theta n_{\cdot} + (y_{\cdot\cdot}) \ln(\theta) - \ln\left(\prod_{i=1}^5 \prod_{j=1}^{n_i} y_{ij}!\right)$$

$$\frac{\partial \ln(L(\theta, y))}{\partial \theta_i} = -n_i + \frac{y_{i\cdot}}{\theta_i} \Rightarrow \hat{\theta}_i = \frac{y_{i\cdot}}{n_i} = \bar{y}_{i\cdot}$$

$$\text{Under } H_0 : \frac{\partial \ln(L(\theta, y))}{\partial \theta} = -n_{\cdot} + \frac{y_{\cdot\cdot}}{\theta} \Rightarrow \tilde{\theta} = \frac{y_{\cdot\cdot}}{n_{\cdot}} = \bar{y}_{\cdot\cdot}$$

$$\Rightarrow \hat{\theta}_1 = \frac{975}{380} = 2.57 \quad \hat{\theta}_2 = \frac{826}{380} = 2.17 \quad \hat{\theta}_3 = \frac{890}{306} = 2.91 \quad \hat{\theta}_4 = \frac{960}{380} = 2.53 \quad \hat{\theta}_5 = \frac{980}{380} = 2.58$$

$$\tilde{\theta} = \frac{975 + 826 + 890 + 960 + 980}{380 + 380 + 306 + 380 + 380} = \frac{4631}{1826} = 2.54$$

$$\frac{\partial^2 \ln(L(\theta, y))}{\partial \theta_i^2} = -\frac{y_{i\cdot}}{\theta_i^2} \quad \frac{\partial^2 \ln(L(\theta, y))}{\partial \theta_i \partial \theta_j} = 0 \quad E\left[\frac{\partial^2 \ln(L(\theta, y))}{\partial \theta_i^2}\right] = -\frac{n_i \theta_i}{\theta_i^2} = -\frac{n_i}{\theta_i}$$

# Testing Equality of Mean Goals Among Countries - II

$$s(\theta, y) = \begin{bmatrix} -n_1 + \frac{y_{1\bullet}}{\theta_1} \\ -n_2 + \frac{y_{2\bullet}}{\theta_2} \\ -n_3 + \frac{y_{3\bullet}}{\theta_3} \\ -n_4 + \frac{y_{4\bullet}}{\theta_4} \\ -n_5 + \frac{y_{5\bullet}}{\theta_5} \end{bmatrix} \quad I(\theta, y) = \begin{bmatrix} \frac{380}{1826\theta_1} & 0 & 0 & 0 & 0 \\ 0 & \frac{380}{1826\theta_2} & 0 & 0 & 0 \\ 0 & 0 & \frac{306}{1826\theta_3} & 0 & 0 \\ 0 & 0 & 0 & \frac{380}{1826\theta_4} & 0 \\ 0 & 0 & 0 & 0 & \frac{380}{1826\theta_5} \end{bmatrix}$$

# Likelihood Ratio Test

$$\begin{aligned}
 \ln\left(L\left(\hat{\theta}, y\right)\right) &= -\sum_{i=1}^5 n_i \hat{\theta}_i + \sum_{i=1}^5 (y_{i\cdot}) \ln\left(\hat{\theta}_i\right) - \ln\left(\prod_{i=1}^5 \prod_{j=1}^{n_i} y_{ij}!\right) \\
 &= -y_{\cdot\cdot} + \sum_{i=1}^5 (y_{i\cdot}) \ln\left(\bar{y}_{i\cdot}\right) - \ln\left(\prod_{i=1}^5 \prod_{j=1}^{n_i} y_{ij}!\right) = \\
 &= -4631 + (918.71 + 641.33 + 950.20 + 889.69 + 928.43) - \ln\left(\prod_{i=1}^5 \prod_{j=1}^{n_i} y_{ij}!\right) \\
 &= -4631 + 4328.36 - \ln\left(\prod_{i=1}^5 \prod_{j=1}^{n_i} y_{ij}!\right) = -302.64 - \ln\left(\prod_{i=1}^5 \prod_{j=1}^{n_i} y_{ij}!\right) \\
 \ln\left(L\left(\tilde{\theta}, y\right)\right) &= -y_{\cdot\cdot} + \ln\left(\tilde{\theta}\right) y_{\cdot\cdot} - \ln\left(\prod_{i=1}^5 \prod_{j=1}^{n_i} y_{ij}!\right) = \\
 &= -4631 + 4309.82 - \ln\left(\prod_{i=1}^5 \prod_{j=1}^{n_i} y_{ij}!\right) = -321.18 - \ln\left(\prod_{i=1}^5 \prod_{j=1}^{n_i} y_{ij}!\right) \\
 \Rightarrow X_{LR}^2 &= -2[-321.18 - (-302.64)] = 37.08 \quad \chi_{4,05}^2 = 9.49
 \end{aligned}$$

Evidence that the true population means differ (in particular: France lower, Germany higher than the others)

# Wald Test

Wald statistic:  $X_w^2 = n \cdot \left( R \hat{\theta} - r \right)^T \left( R I^{-1} \left( \hat{\theta} \right) R^T \right)^{-1} \left( R \hat{\theta} - r \right)$

$$R \hat{\theta} - r = \begin{bmatrix} 1 & -1 & 0 & 0 & 0 \\ 1 & 0 & -1 & 0 & 0 \\ 1 & 0 & 0 & -1 & 0 \\ 1 & 0 & 0 & 0 & -1 \end{bmatrix} \begin{bmatrix} 2.57 \\ 2.17 \\ 2.91 \\ 2.53 \\ 2.58 \end{bmatrix} - \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 2.57 - 2.17 - 0 \\ 2.57 - 2.91 - 0 \\ 2.57 - 2.53 - 0 \\ 2.57 - 2.58 - 0 \end{bmatrix} = \begin{bmatrix} 0.40 \\ -0.34 \\ 0.04 \\ -0.01 \end{bmatrix}$$

$$R I^{-1} \left( \hat{\theta} \right) R^T = \begin{bmatrix} 1 & -1 & 0 & 0 & 0 \\ 1 & 0 & -1 & 0 & 0 \\ 1 & 0 & 0 & -1 & 0 \\ 1 & 0 & 0 & 0 & -1 \end{bmatrix} \begin{bmatrix} \frac{1826(2.57)}{380} & 0 & 0 & 0 & 0 \\ 0 & \frac{1826(2.17)}{380} & 0 & 0 & 0 \\ 0 & 0 & \frac{1826(2.91)}{306} & 0 & 0 \\ 0 & 0 & 0 & \frac{1826(2.53)}{380} & 0 \\ 0 & 0 & 0 & 0 & \frac{1826(2.58)}{380} \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 & 1 \\ -1 & 0 & 0 & 0 \\ 0 & -1 & 0 & 0 \\ 0 & 0 & -1 & 0 \\ 0 & 0 & 0 & -1 \end{bmatrix}$$

$$= 1826 \begin{bmatrix} .0068 & -0057 & 0 & 0 & 0 \\ .0068 & 0 & -0.0095 & 0 & 0 \\ .0068 & 0 & 0 & -0.0067 & 0 \\ .0068 & 0 & 0 & 0 & -0.0068 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 & 1 \\ -1 & 0 & 0 & 0 \\ 0 & -1 & 0 & 0 \\ 0 & 0 & -1 & 0 \\ 0 & 0 & 0 & -1 \end{bmatrix} = 1826 \begin{bmatrix} .0125 & .0068 & .0068 & .0068 \\ .0068 & .0163 & .0068 & .0068 \\ .0068 & .0068 & .0135 & .0068 \\ .0068 & .0068 & .0068 & .0136 \end{bmatrix}$$

$$\Rightarrow \left( R I^{-1} \left( \hat{\theta} \right) R^T \right)^{-1} = \frac{1}{1826} \begin{bmatrix} 132.72 & -25.34 & -36.23 & -35.49 \\ -25.34 & 89.96 & -21.80 & -21.36 \\ -36.23 & -21.80 & 119.25 & -30.53 \\ -35.49 & -21.36 & -30.53 & 117.44 \end{bmatrix} \Rightarrow X_w^2 = 38.33$$

# Lagrange Multiplier (Score) Test

Lagrange Multiplier (Score) Statistic:  $X_{LM}^2 = \frac{1}{n_{\bullet}} s(\tilde{\theta}, y)^T \left( I(\tilde{\theta}) \right)^{-1} s(\tilde{\theta}, y)$

$$\tilde{\theta} = \bar{y}_{\bullet\bullet} = \frac{4631}{1826} = 2.5361$$

$$s(\tilde{\theta}, y) = \begin{bmatrix} -380 + \frac{975}{\tilde{\theta}} \\ -380 + \frac{826}{\tilde{\theta}} \\ -306 + \frac{890}{\tilde{\theta}} \\ -380 + \frac{960}{\tilde{\theta}} \\ -380 + \frac{980}{\tilde{\theta}} \end{bmatrix} = \begin{bmatrix} 4.42 \\ -54.31 \\ 44.93 \\ -1.47 \\ 6.41 \end{bmatrix} \quad I(\tilde{\theta}, y) = \begin{bmatrix} \frac{380}{1826\tilde{\theta}} & 0 & 0 & 0 & 0 \\ 0 & \frac{380}{1826\tilde{\theta}} & 0 & 0 & 0 \\ 0 & 0 & \frac{306}{1826\tilde{\theta}} & 0 & 0 \\ 0 & 0 & 0 & \frac{380}{1826\tilde{\theta}} & 0 \\ 0 & 0 & 0 & 0 & \frac{380}{1826\tilde{\theta}} \end{bmatrix}$$

$$\Rightarrow s(\tilde{\theta}, y) = \begin{bmatrix} 4.42 \\ -54.31 \\ 44.93 \\ -1.47 \\ 6.41 \end{bmatrix} \Rightarrow I(\tilde{\theta}, y)^{-1} = \begin{bmatrix} 12.19 & 0 & 0 & 0 & 0 \\ 0 & 12.19 & 0 & 0 & 0 \\ 0 & 0 & 15.13 & 0 & 0 \\ 0 & 0 & 0 & 12.19 & 0 \\ 0 & 0 & 0 & 0 & 12.19 \end{bmatrix}$$

$$\Rightarrow X_{LM}^2 = 36.83$$

# Testing Goodness of Fit to Poisson Distribution

- All estimation and testing has assumed that number of goals follow Poisson distributions
- To test whether that assumption is reasonable, we compare the observed distributions of goals with what we would expect under the Poisson model
- We can check whether the observed mean and variance are similar (under Poisson model they are equal)
- We can also obtain a chi-square statistic by summing over range of goals:  $(\text{observed\#} - \text{expected\#})^2 / \text{expected\#}$  which under hypothesis of model fits is approximately chi-square with  $(\# \text{ in range}) - 1$  degrees of freedom

# Distributions of Goals

Goals	Observed					Expected (Truncated at 7)					Chi-Square Statistic					
	England	France	Germany	Italy	Spain	England	France	Germany	Italy	Spain	England	France	Germany	Italy	Spain	
0	30	54	18	36	29	29.2062	43.2279	16.6947	30.3822	28.8244	0.0216	2.6843	0.1021	1.0388	0.0011	
1	79	82	43	85	73	74.9370	93.9639	48.5563	76.7549	74.3367	0.2203	1.5233	0.6358	0.8857	0.0240	
2	99	110	66	85	96	96.1363	102.1239	70.6130	96.9536	95.8553	0.0853	0.6074	0.3014	1.4738	0.0002	
3	67	57	77	78	79	82.2218	73.9950	68.4592	81.6451	82.4019	2.8180	3.9034	1.0655	0.1627	0.1404	
4	61	51	54	49	60	52.7410	40.2105	49.7783	51.5653	53.1275	1.2933	2.8951	0.3580	0.1276	0.8890	
5	24	15	29	20	28	27.0645	17.4810	28.9560	26.0541	27.4026	0.3470	0.3521	0.0001	1.4068	0.0130	
6	11	4	13	20	8	11.5736	6.3330	14.0364	10.9701	11.7783	0.0284	0.8595	0.0765	7.4328	1.2120	
7	6	6	4	4	6	6.1196	2.6648	8.9060	5.6747	6.2732	1.3558	7.0529	0.9482	0.3095	0.0842	
8	2	1	1	1	1	#N/A	#N/A	#N/A	#N/A	#N/A						
9	1	0	1	1	0	#N/A	#N/A	#N/A	#N/A	#N/A						
10	0	0	0	1	0	#N/A	#N/A	#N/A	#N/A	#N/A						
Total Games	380	380	306	380	380						Chi-square	6.1697	19.8780	3.4876	12.8377	2.3640
Total Goals	975	826	890	960	980						CritVal	14.0671	14.0671	14.0671	14.0671	14.0671
Average	2.5658	2.1737	2.9085	2.5263	2.5789						P-Value	0.5201	0.0058	0.8365	0.0762	0.9370

$$X_{obs}^2 = \sum_{i=0}^7 \frac{(\text{obs}_i - \text{exp}_i)^2}{\text{exp}_i} \approx \chi_7^2$$

All leagues, except France, appear to be well described by the Poisson distribution. Especially England, Germany, and Spain