

# Stat13-lecture 25

## regression (continued, SE, t and chi-square)

- Simple linear regression model:
- $Y = \beta_0 + \beta_1 X + \epsilon$
- Assumption :  $\epsilon$  is normal with mean 0 variance  $\sigma^2$

The fitted line is obtained by minimizing the sum of squared residuals; that is finding  $\beta_0$  and  $\beta_1$  so that

$(Y_1 - \beta_0 - \beta_1 X_1)^2 + \dots + (Y_n - \beta_0 - \beta_1 X_n)^2$  is as small as possible

- This method is called least squares method

# Least square line is the same as the regression line discussed before

- It follows that estimated slope  $\hat{\beta}_1$  can be computed by
- $r [SD(Y)/SD(X)] =$   
 $[cov(X,Y)/SD(X)SD(Y)][SD(Y)/SD(X)]$
- $= cov(X,Y)/VAR(X)$  (this is the same as equation for  $\hat{\beta}_1$  on page 518)
- The intercept  $\hat{\beta}_0$  is estimated by putting  $x=0$  in the regression line; yielding equation on page 518
- Therefore, there is no need to memorize the equation for least square line; computationally it is advantageous to use  $cov(X,Y)/var(X)$  instead of  $r[SD(Y)/SD(X)]$

# Finding residuals and estimating the variance of $\epsilon$

- Residuals = differences between  $Y$  and the regression line (the fitted line)
- An unbiased estimate of  $\epsilon^2$  is
- $[\text{sum of squared residuals}] / (n-2)$
- Which divided by  $(n-2)$  ?
- Degree of freedom is  $n-2$  because two parameters were estimated
- $[\text{sum of squared residuals}] / \epsilon^2$  follows a chi-square.

# Hypothesis testing for slope

- Slope estimate  $\hat{\beta}_1$  is random
- It follows a normal distribution with mean equal to the true  $\beta_1$  and the variance equal to  $\sigma^2 / [n \text{ var}(X)]$

Because  $\sigma^2$  is unknown, we have to estimate from the data ; the SE (standard error) of the slope estimate is equal to the squared root of the above

# t-distribution

- Suppose an estimate  $\hat{\mu}$  is normal with
- variance  $c \sigma^2$ .
- Suppose  $\sigma^2$  is estimated by  $s^2$  which is related to a chi-squared distribution
- Then  $(\hat{\mu} - \mu) / \sqrt{(c s^2)}$  follows a t-distribution with the degrees of freedom equal to the chi-square degree freedom

# An example

- Determining small quantities of calcium in presence of magnesium is a difficult problem of analytical chemists. One method involves use of alcohol as a solvent.
- The data below show the results when applying to 10 mixtures with known quantities of CaO. The second column gives
- Amount CaO recovered.
- Question of interest : test to see if intercept is 0 ; test to see if slope is 1.

X:CaO present	Y:CaO recovered	Fitted value	residual
4.0	3.7	3.751	-.051
8.0	7.8	7.73	.070
12.5	12.1	12.206	-.106
16.0	15.6	15.688	-.088
20.0	19.8	19.667	.133
25.0	24.5	24.641	-.141
31.0	31.1	30.609	.491
36.0	35.5	35.583	-.083
40.0	39.4	39.562	-.161
40.0	39.5	39.562	-.062

## Least Squares Estimates:

	Estimate	Standard error
• Constant	-0.228090	(0.137840)
• Predictor	0.994757	(5.219485E-3)
• R Squared:	0.999780	Squared correlation
• Sigma hat:	0.206722	Estimate of
• Number of cases:	10	SD( $\square$ )
• Degrees of freedom:	8	

$$.22809 / .1378 = 1.6547$$

$$(1 - 0.994757) / 5.219485E-3 = 1.0045052337539044$$