

A.

1. if residuals follows ~~uniform~~ ^{distribution} with mean zero ~~and~~ uniform
2. the number of variables whose value can be determined freely. it's like the dim. of vector space.
3. dividing ^{error residuals} by the population sd ($e_{d,i}$) vs. by the sample sd ($e_{s,i}$)
4. To penalize model whose independent variables are too many. R^2 increases even if the model include variables with less explanatory power.
5. Assuming normality is essential.
6. No. only when $\delta^2 = \delta_1^2 = \delta_2^2$
7. No.

B.

1. chi-squared distribution with ~~df=1~~ ^{df=1}
2. F-distribution with $df=(2,2)$
- 3.
- 4.
5. F-distribution with $df=(1,1)$

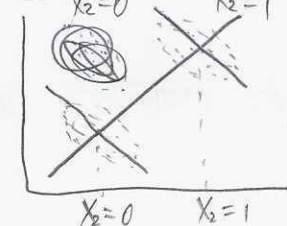
C. linear model should be linear about $\beta_0, \beta_1, \dots, \beta_k$. 2, 3, 5.

D.

1. (consistent)



(inconsistent)



2. No. The interaction effect could made unexpected main effect. models only use X_1 ($Y_i = \beta_0 + \beta_1 X_{1i} + \epsilon_{1i}$) or X_2 ($Y_i = \beta_0 + \beta_1 X_{2i} + \epsilon_{2i}$) can show significant marginal effect.

E.

1. (women) $E(y_i) = \beta_0 \neq \mu_2$
(men) $E(y_i) = \beta_0 + \beta_1 = \mu_1$

$\therefore \beta_0 = \mu_2 \quad \beta_1 = \mu_1 - \mu_2$

$$\hat{\beta}_1 = \frac{S_{xy}}{S_{xx}} = \frac{-\frac{n_1 n_2 \sum y_i + \sum y_i}{n_1 + n_2}}{\frac{n_1 n_2}{n_1 + n_2}} = \frac{n_1 \mu_1 - \frac{n_1}{n_2} (\mu_1 + \mu_2)}{\frac{n_1 n_2}{n_1 + n_2}}$$

$$\hat{\beta}_0 = \bar{y} - \hat{\beta}_1 \bar{x} = \bar{y} - \frac{n_1 \mu_2}{n_2} \left(\frac{-\frac{n_1 \mu_2}{n_1 n_2}}{\frac{n_1 n_2}{n_1 + n_2}} \right)$$
$$= \frac{n_1 \mu_1 + \mu_2 n_2}{n_1 + n_2} - \frac{n_1 \mu_2}{n_2}$$

2. $\beta_1 = 0$ equals to $\mu_1 - \mu_2 = 0 (\Leftrightarrow \mu_1 = \mu_2)$

test statistic for

$$\left(\begin{aligned} \beta_1 = 0 : t &= \frac{\hat{\beta}_1}{se(\hat{\beta}_1)} = \frac{\hat{\beta}_1}{\sqrt{\frac{\hat{\sigma}^2}{S_{xx}}}} = \frac{\hat{\beta}_1}{\sqrt{\hat{\sigma}^2 \left(\frac{1}{n_1} + \frac{1}{n_2} \right)}} \\ \mu_1 = \mu_2 : t &= \frac{(\mu_1 - \mu_2) - 0}{\sqrt{\hat{\sigma}^2 \left(\frac{1}{n_1} + \frac{1}{n_2} \right)}} \end{aligned} \right)$$

$$S_{xx} = n_1 \left(1 - \frac{n_1}{n_1 + n_2} \right)^2 + n_2 \left(-\frac{n_1}{n_1 + n_2} \right)^2 = \frac{n_1 n_2}{n_1 + n_2}$$
$$= \frac{1}{\frac{1}{n_1} + \frac{1}{n_2}}$$

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\therefore Two statistics are same.

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F.

G.

H.

I. $100 \leq X_i, Y_i \leq 320$,

1. $Y_i = 100 + \beta_1 X_i + \varepsilon_i$

2. for model $Y_i = \beta_0 + \beta_1 X_i + \varepsilon_i$,
reduced model is ($\beta_1 | \beta_0 = 100$):
 $Y_i = 100 + \beta_1 X_i + \varepsilon_i$ but ~~the~~ intercept
is about $Y_i | X_i = 0$, which means intercept
should be lower than 100 because $100 \leq X_i \leq 320$

3. $\sum (Y_i - \beta_0 - \beta_1 X_i)^2$ should be
minimized.

J.1. Adjusted $R^2 = 1 - (1 - 0.938) \times \frac{99}{97}$
($K=2, n=100$)
 $= 1 - 0.062 \times \frac{99}{97}$
 $1 - \frac{SSE}{SST} \times \frac{n-1}{n-K-1}$

2. $\frac{SSR}{0.9428/97} = 733.5$, $SSR = \frac{2 \cdot 0.9428 \cdot 733.5}{97}$

3. From F-test with $df = (2, 97)$,
 $F = \frac{SSR/2}{SSE/97} = 733.5$ follows ~~F~~

P-value for H_0 : less than $2.2e^{-16}$.

4. Statistically, addition of x_3 isn't
significant because R^2 didn't
show any significant increase (to put it
another way, Adjusted R^2 decreased),
and F-statistic for reduced model
($\beta_0, \beta_1, \beta_2 | \beta_3$) is greater than the full
model ($\beta_0, \beta_1, \beta_2, \beta_3$)