FC402 $(2023124)$
Ragur's Speakng Noifs For Cass
$\qquad$

$$
\text { Vassihs } 184\left(W k 5, A_{T}\right)
$$

Plan

$$
\begin{aligned}
& \left(Q_{1}\right) \\
& \left(Q_{2}\right) \\
& \left(Q_{3}\right)
\end{aligned}
$$

(MAYBE: FWL Theorem if WE have time. otherwise SEE MY Notes on webpage)

PS 6

$$
\text { QUa) }(a) h_{i}=\beta_{0}+\sum_{j=1}^{4} \beta_{j} d_{j i}+x_{\left.(1 x k)(k x)^{\prime}\right)}^{\varepsilon_{i}} f_{0} i=1, \ldots, N
$$

Define: $Q_{N \times(k+5)}=\left[i \quad d_{1} d_{2} d_{3}^{(1 \times x)(k x x)} d_{4} X\right]$. Clearly, $i=\sum_{j=1}^{4} d_{j}$ since every individual belongs in $\mid$ and only 1 category (by design).
$\therefore$ Fat least I perfect lines relationship among the regressor $\Rightarrow A 1$ is violated!
Define: $\underset{(k+5) \times 1}{B}=\left[\beta_{0}, \ldots, \beta_{4}, \gamma^{\prime}\right]^{\prime}$. Then, $\hat{B}_{0, S}=\left(Q^{\prime} Q\right)^{-1} Q^{\prime} h$ cannot be estimated since rank $\left(Q^{\prime} Q\right)<(k+5)$ and $\left.Q^{\prime} Q\right)^{\prime}$ will not exist.
This is a classic instance of the "dummy variable trap".
(b) 2 possible ways to avoid the aforesaid "trap":
(i) $h_{i}=\beta_{0}+\sum_{j=2}^{L} \beta_{j} d_{j i}+x_{i}^{\prime} \gamma+\varepsilon_{i}$ for $i=I_{1} \ldots, N \rightarrow$ drop "dI"

Here, "individuals with guarabled income lear I" becomes the reference or benchmark Category.
$E[h \mid Q]=\beta_{0}+x_{\gamma}$ for this reference category, then for $j=2, \ldots, 4$,
$\beta_{j}$ is the marginal/additional effect (note: not necessarily positive!) on the conditional mean of $h$ for individuals belonging in the $j$ th guaranteed income level group; that is, over and above that for individuals in the reference category.

Note: $\beta_{0}$ is the intercept for the reference category

$$
\begin{array}{ll}
\beta_{0}+\beta_{2} & -1-\operatorname{gropp} 2 \\
\beta_{0}+\beta_{3} & -1-g r o p \\
\beta_{0}+\beta_{4} & -1-g r o p
\end{array} 4
$$

(ii) $h_{i}=\sum_{j=1}^{4} \theta_{j} d_{j i}+x_{i}^{\prime} \delta+幺$ for $i=1, \ldots, N \rightarrow$ drop the Constant

Here, we no longer have a reference category.
ie. The interests by category ard simply $0_{j}$ for $j=1, \ldots, 4$ respectively.
[Tip: in practice, specification (i) is more commonly used. The choice of reference category is arbitrary and usually based on the question of interest to the lesearchel.]
(iii) How are the coefficients related?

$$
\theta_{1}=\beta_{0} ; \theta_{2}=\beta_{0}+\beta_{2} ; \theta_{3}=\beta_{0}+\beta_{3} ; \theta_{4}=\beta_{0}+\beta_{4} ; \delta=\gamma
$$

The OIS residuals should be identical since $\hat{h}=\hat{Q}$ ours will be identical in both regressions.

QL) $y=s \beta_{1}+x_{2} \beta_{2}+\varepsilon$; Note: $x_{2}$ cannot contain a constant!

$$
M_{S}:=\left[I_{T}-S\left(s^{\prime} s^{-1} s^{\prime}\right]_{\substack{\text { where }}} S=\left[\begin{array}{llll}
d_{1} & d_{2} & d_{3} & d_{4}
\end{array}\right]\right.
$$

Consider $z=\delta_{1} d_{1}+\delta_{2} d_{2}+\delta_{3} d_{3}+\delta_{4} d_{4}+\eta ;\left[S^{\prime} s \int_{i j}=d_{i}^{\prime} d_{j}\right.$ for $i, j=1_{1}, \ldots 4$

- note the that $d_{i}^{\prime} d_{j}=\left\{\begin{array}{c}T_{i} \text { if } i=j \text { so that }\left(S^{\prime} S\right)=\operatorname{diag}\left\{T_{1}, \ldots, T_{4}\right\} \\ 0 \\ \text { olw }\end{array}\right\}$ and $\left(S^{\prime} S\right)^{-1}=\operatorname{dian}\left\{\frac{1}{T_{1}}, \ldots, \frac{1}{T_{4}}\right\}$
Notation:
- Further, $S_{z}^{\prime}=\left[\begin{array}{l}d_{1}^{\prime} z \\ d_{2}^{\prime} z \\ d_{3}^{\prime} z \\ d_{4}^{\prime} z\end{array}\right]=\left[\begin{array}{c}\sum_{i=1}^{T} z_{i} \mathbb{1}(Q 1)_{i} \\ \sum_{i=1}^{1} z_{i} \mathbb{1}(Q 2)_{i} \\ \sum_{i=1}^{1} z_{i} \mathbb{1}(Q 3)_{i} \\ \sum_{i=1}^{\sum} z_{i} \mathbb{1}(Q 4)_{i}\end{array}\right]=\left[\begin{array}{l}\sum_{i=1}^{T} z 1_{i} \\ \sum_{i=1}^{2} z 2_{i} \\ \sum_{i=1}^{2} z_{i} \\ \sum_{i=1}^{4} z 4_{i}\end{array}\right]$
- Thus $\hat{\delta}=\left(S S^{-1} S^{\prime} z=\left[\bar{z}_{1}, \ldots, \overline{z_{4}}\right]^{\prime}\right.$
- So, $S \hat{\delta}$ yields a $T x \mid$ matrix st. for any

- The darned structure of $M_{S} z=z-S \hat{\delta}$ follows.
(b) Consider the descafonalised vaiables regrestion:
$y^{*}=x^{*} \alpha_{2}+\varepsilon^{*}$, where * denotes pre-multiplication by $M_{S}$.
Then, $\hat{\alpha}_{\text {ois }}=\left(x^{+\prime} x^{-1}\right) x^{* \prime} y^{*}=\left(\left(M_{s} x\right)^{\prime} M_{s} x^{-1}\left(M_{s} x\right)^{\prime} M_{s} y=\left(x^{\prime} M_{s} x^{-1} x^{\prime} M_{s} y=\hat{\beta}_{\text {ols }}\right.\right.$ by the Ful

Dear all,
A student in one class asked me a question about the DV trap. The question was roughly as follows:

Consider $y=\alpha+\beta_{M} M+\beta_{F} F+\varepsilon$ where $M$ and $F$ are Male and female dummies. Obviously, this specification suffers for the DV trap since $i=M+F$ so either we drop $i$ or M or $F$. The student asked that if we drop $i$, we still have a linear relationship in the suse that $M+F=1$. So why is this not a problem?

Math:
(1) obviously, $x:=[i M]$ is not a problem since $i$ and $M$ are not collinear.

The rack of $x$ is full and $\left(x^{\prime} x\right)^{-1}$ can be computed.
(2). But this is identical to a regression of $y_{\text {an }} X:=[M F]$. To seethes:

$$
\text { consides } \begin{aligned}
y & =\delta_{M} M+\delta_{F} F+\varepsilon \\
& =\delta_{M} M+\delta_{F}(1-M)+\varepsilon \\
& =\delta_{M} M+\delta_{F}-\delta_{F} M+\varepsilon \\
& =\delta_{F}+\left(\delta_{M}-\delta_{F}\right) M+\varepsilon \\
& =\theta_{0}+\frac{\theta_{1}}{} M+\varepsilon
\end{aligned}
$$

$\therefore$ since the two specifications are effectively the same if you are happy with (1) you must also be happy with (2).'

TUITION:
I think the best way to appooach this problem is to think directly in terms of rank $\left(x^{\prime} x\right)$ since that is the precise Condition to check for AI. When $x:=\left[\mathcal{C M F}\right.$, rank of $x^{\prime} x$ will be 2 but also when $X:=[M F)$ and when $X:=(i M)$ and when $X:=[i F]$.

Hope that helps,
Raguir

Q3. $Z \in \mathbb{R}^{n}, z \sim\left(\mu_{z}, V_{z}\right)$

- constant $q \in \mathbb{R}^{n}$
(a) Define $\omega:=q^{\prime} z$. Find $E(w) \& \operatorname{Var}(\omega)$.

$$
\begin{aligned}
& E(\omega)=q^{\prime} E(z)=q^{\prime} \mu_{z} . \\
& \operatorname{Var}(\omega)=q^{\prime} \operatorname{Var}(z) q=q^{\prime} \operatorname{V}_{z} q .
\end{aligned}
$$

(b) Suppose that for a particular choice of $q \neq 0, \operatorname{Var}(w)=0$. What does this say
(i) about the random vector $z$ ? What does it say about the rank of the variancecovariance matrix $V_{z}$ ?
(ii) ${ }^{1}$
(i) Since there exists a $q \neq 0$ s.t. $\operatorname{Var}(q z z)=0$, there exists a perfect linear relationship between the elements of $z$.

You don't reed to read this (sketch of a) proof. Last year students wanted me to give extra intuition for the second paragraph of Vassilis' Sowtion to I cane up with the reasoning below. That is all this is!
(ii) let's assume WLOG that there exists just one vector $q \neq 0$ s.t. $\operatorname{Var}\left(q^{\prime} z\right)=0$.

Men, recognise that

1. $\operatorname{Var}\left(q^{\prime} z\right)=q^{\prime} V_{z} q=0$ if $V_{z} q=0$ for $q \neq 0$;
2. $\operatorname{Null}\left(V_{z}\right):=\left\{q \in \mathbb{R}^{n}: V_{z q}=0\right\}$ and so

$$
\operatorname{Nullity}\left(V_{z}\right):=\operatorname{dim}\left\{\operatorname{Null}\left(V_{z}\right)\right\}=1 \text {; }
$$

3. By the rak-nullity theorem,

$$
\operatorname{ruk}\left(V_{z}\right)+\operatorname{Nullity}\left(V_{2}\right)=n
$$

$\therefore B y\left(1,3, \operatorname{rank}\left(v_{z}\right)=n-1\right.$.

ATTEMPT 1
Let's see if I can think of an example for you for $Q 3$ :
Consider $x \sim N\left(0,0^{2}\right)$ and the vector $(x, 2 x)^{\prime}$.
Then $\binom{x}{2 x}$ has variance $\left(\begin{array}{ll}\sigma^{2} & 2 r^{2} \\ 2 \sigma^{2} & 4 \sigma^{2}\end{array}\right)$, right?
Now ask yourself what is the rate of the matrix above?!


