

2016 Q1

$$- y_i \sim \text{Pois}(\lambda_i), \quad \lambda_i = x_i' \beta$$

$$P(y_i = y) = \frac{\lambda^y \exp(-\lambda)}{y!}, \quad y \geq 0.$$

$$\Rightarrow E[y] = \text{Var}(y) = \lambda$$

1.1)

$$y_i = x_i' \beta + \varepsilon_i \text{ (vector notation)}$$

Just the OLS estimator

$$\hat{\beta} = E_n[x_i x_i']^{-1} E_n[x_i y_i]$$

which we know is consistent if $E[x_i x_i']$ has full rank and

$$\begin{aligned} E[\varepsilon_i | x_i] &= E[y_i - x_i' \beta | x_i] \\ &= E[y_i | x_i] - E[x_i' \beta | x_i] \\ &= 0. \end{aligned}$$

12) $\hat{\beta}$ unbiased?

- $\hat{\beta}$ unbiased by property of OLS
(since $E[\varepsilon_i | x_i] = 0$).
- $\hat{\beta}$ linear estimator as req. by Gauss-Markov.
- Not BLUE because our problem has
 $E[y_i | x_i] = \text{Var}[y_i | x_i] = \lambda_i$

Recall, to use G-M Thm:

- 1) $\hat{\beta}$ unbiased.
 $\hookrightarrow E[\hat{\beta} | X] = \beta$
- 2) $\text{var}(\hat{\beta} | X) = \sigma^2 (X'X)^{-1}$

However,

$$\begin{aligned}\lambda_i &= \text{var}(y_i | x_i) = \text{var}(x_i' \beta + \varepsilon_i | x_i) \\ &= \text{var}(\varepsilon_i | x_i) \\ &= \lambda_i \\ &= x_i' \beta\end{aligned}$$

1.3) Assuming $\pi_i = \exp(x_i'\beta)$ true

$\hookrightarrow \log y_i = x_i'\beta + \varepsilon_i$ can we do this?

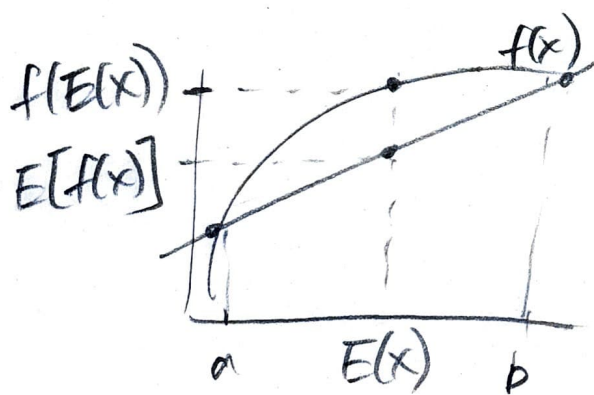
Before!

If $\pi_i = \exp(x_i'\beta)$, then

$$E[y_i | x_i] = \pi_i = \exp(x_i'\beta) \quad \text{"true"}$$

After.

$$\begin{aligned} E[\log y_i | x_i] &\leq \log(E(y_i | x_i)) \\ &= \log(\exp(x_i'\beta)) \\ &= x_i'\beta \end{aligned}$$



Not true by Jensen's
Inequality

1.4) Estimate

$$y_i = x_i' \beta + \varepsilon, \text{ where } y_i \sim \text{Poiss}(\pi_i)$$

Goal: $\hat{\beta}$

$$\pi_i = \exp(x_i' \beta)$$

$$\pi_i = \exp(x_i' \beta)$$

$$\text{Recall, } \underline{E[y_i]} = \text{Var}(y_i) = \pi_i = \underline{\exp(x_i' \beta)}.$$

We can perform a nonlinear GMM estimation with moment condition.

$$E[y_i - \underbrace{\exp(x_i' \beta)}_{E(y_i) = \pi_i}] = 0$$

2024 Q2

- assume $E[\varepsilon^2|X] = \sigma^2 I_n$.

2.1)

1) $X = \delta Z + \varepsilon$

2) $Y = \hat{\beta} \hat{X}$

\Rightarrow The $\hat{\beta}$ is $\hat{\beta}_{OLS}$ (or $\hat{\beta}_{IV}$ if $l=k$)

Requires:

- homoscedasticity
- $l \geq k$
- $E[Z\varepsilon] = 0$

2.2)

1) $\hat{X} = \delta Z$

1.5) $X = \hat{X} + \hat{\eta}$

2) $Y = \beta_2(\hat{\eta}) + \varepsilon, \hat{\eta} = X - \hat{X}$

$\hat{\eta}$ = amount of variation \hat{X} cannot explain in X

Stage (2) is equivalent to $Y = \beta_0 + \underbrace{\phi_1 \hat{X}}_{\text{}} + \beta_2 X + e$

$$= \underbrace{\phi_1 \delta}_{\text{}} Z$$

$$= \beta_1 Z$$

Hence, this is just a typical multivariate OLS model where we require

- $l \geq k$
- $E[X\varepsilon] = E[Z\varepsilon] = 0$.

Note: This is a full depiction of OLS.

2.3) Obviously, if the assumptions hold for both models, either estimator works.

If we have that $E[Z\varepsilon] = 0$, but $E[X\varepsilon] \neq 0$
 \Rightarrow IV.

However, if our goal is to simply find the causal effect of X on Y (assuming all assumptions hold), we could perform simple OLS.

Side Note: Could perform SUR if $E[Z\varepsilon] = 0$ and $l \geq k$.

2014 Q3

3.1)

$$\hat{\theta}(\hat{W}) = (S' \hat{W} S)^{-1} S' \hat{W} s$$

$$\Rightarrow \hat{\beta}(\hat{W}) = (X' Z \hat{W} Z' X)^{-1} (X' Z \hat{W} Z' Y)$$

$$S = Z' X$$

$$s = Z' Y$$

Note: There doesn't seem to be instruments here, so $Z = X$

$$\theta = \begin{bmatrix} \alpha \\ \beta \\ \delta \end{bmatrix}$$

$$S = E_n[X' X]$$

$$s = E_n[X' Y] = E_n \begin{bmatrix} x_i \varepsilon_i \\ x_i \eta_i \\ x_i y_i \end{bmatrix}$$

For $\hat{W} \xrightarrow{P} W$, where W is the efficient weighting matrix (i.e. true inverse of $\text{var}(\varepsilon)$), then my \hat{W} must at least be symmetric and positive definite.

3.2) Here, we have a just-identified model ($l=k$), so the choice of \hat{w} does not matter because things will cancel out.

Hence, estimating this model eq-by-eq vs. jointly are identical (assuming w is full rank).

3.3) WLOG, if $\alpha' = 0$, then we have an overidentified model in which

$$l = 3 > k = 2.$$

Hence, we should estimate jointly to attain better efficiency.

However, the drawback of estimating jointly is that misspecification will permeate throughout the model, so other estimators will be affected as well.