Econometrics I - TA section

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Random variables

A random variable is a real-valued outcome; a function from the sample space S i.e. outcomes to the real line \mathcal{R} .

- Can be discrete i.e. a dice roll
- Can be cts i.e. a stock price

Cumulative distribution function

The distribution function is $F(x) = P[X \le x]$, the probability of the event $\{X \le x\}$.

- 1. F(x) is non-decreasing
- 2. $F(x) \ge 0$
- 3. $F(x) \le 1$

Probability density function When F(x) is differentiable, its density is

•
$$f(x) = \frac{d}{dx}F(x)$$

•
$$F(x) = \int_{-\infty}^{x} f(t)dt$$

A function f(x) is a density function if and only if

- 1. $f(x) \ge 0$ for all x.
- $2. \int_{-\infty}^{\infty} f(x) dx = 1$

Expectations

- Discrete random variable $E(X) = \sum_{x \in S} x Pr(X = x)$
- $E(x) = \int f(x) dx$

Variance

- $Var(X) = E((X E(x))^2)$
- $Var(X) = E(X^2) E(X)^2$

2.2 HIE

X is uniformly distributed. Find the cdf of $Y = \log(X/1-X)$ In this case, Y = g(X)

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For the uniform distribution,

$$F_{x}(x) = x$$

where $0 \le x \le 1$ Thus,

$$F_y(y) = g^{-1}(y) = e^y/(e^y - 1)$$

2.6 HIE Find E(X) and Var(X).

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 $E(X^2) = \int_0^1 ax^{a-1}x^2 = \frac{a}{2+a}x^{2+a}|_0^1 = \frac{a}{2+a}$

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 $E(X) = \sum_{x=1}^{n} \frac{x}{n} = \frac{1}{n} \sum_{x=1}^{n} x = \frac{1}{n} \left(\frac{n(n+1)}{2} \right) = (n+1)/2$
 $E(X^2) = \sum_{x=1}^{n} \frac{x}{n} = \frac{1}{n} \sum_{x=1}^{n} x^2 = \frac{1}{n} \left(\frac{n(n+1)(2n+1)}{6} \right) = \frac{(n+1)(2n+1)}{6}$

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 $E(X) = \int_0^2 3/2(x-1)^2 x dx = 1$
 $E(X^2) = \int_0^2 3/2(x-1)^2 x^2 dx = \frac{8}{5}$

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$$f_{y}(y) = \left| \frac{dg^{-1}(y)}{dy} \right| f_{x}(g^{-1}(y))$$

For this problem, $g^{-1}(y) = e^{-y}$ and $\frac{dg^{-1}(y)}{dy} = -e^{-y}$ Thus

$$f_y(y) = e^{-e^{-y}}e^{-y}$$

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2.14 HIE Show that if X is a continuous random variable, then min E|X-a|=E|X-m| where m is the median Split the integral up into 2 sets $\{X< a\}$ and $\{X\geq a\}$

$$\int_{-\infty}^{a} (a-x)f(x)dx + \int_{a}^{\infty} (x-a)f(x)dx$$

Take derivative of this expression, to find the maximum. Use Leibniz integral rule

$$(a-x)f(x)|_{x=a} + \int_{-\infty}^{a} f(x)dx + (x-a)f(x)|_{x=a} - \int_{a}^{\infty} f(x)dx$$
$$= \int_{-\infty}^{a} f(x)dx - \int_{a}^{\infty} f(x)dx = 0$$

Thus F(a) = 1 - F(a) and F(a) = .5 as required.

3.2 For the Binomial distribution show

$$\sum_{x=0}^n \pi(x;n,p) = 1$$

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$$\sum_{x=0}^{n} {n \choose x} p^{x} (1-p)^{n-x} =$$

$$([1-p] + [p])^{n} = 1$$

From the binomial theorem

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$$\sum_{x=0}^{n} x \binom{n}{x} p^{x} (1-p)^{n-x}$$

Factoring out an np, this gives:

$$= np \sum_{x=1}^{n} \frac{(n-1)!}{(n-x)!(x-1)!} p^{x-1} (1-p)^{n-x}$$

Can apply the binomial theorem to

$$np\sum_{x=1}^{n} {n-1 \choose x-1} p^{x-1} (1-p)^{n-x} = np(p+[1-p])^{n-1} = np$$

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$$Var(X) = np(1-p) = np - np^2$$

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Need to calculate $E(X^2)$

$$\sum_{x=0}^{n} x^2 \binom{n}{x} p^x (1-p)^{n-x}$$

Can use a similar factoring trick

$$= n(n-1)p^{2} \sum_{x=2}^{n} {n-2 \choose x-2} p^{x-2} (1-p)^{n-x} + np \sum_{x=1}^{n} {n-1 \choose x-1} p^{x-1} (1-p)^{n-x}$$

Can apply the binomial theorem to get

$$E(X^2) = n(n-1)p^2 + np$$

Hence

$$Var(X) = E(X^2) - E(X)^2 = n(n-1)p^2 + np - [(np)^2] = np(1-p)$$

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$$\sum_{x=0}^{\infty} \frac{\lambda^x e^{-\lambda}}{x!} = e^{-\lambda} \sum_{x=0}^{\infty} \frac{\lambda^x}{x!}$$

Turns out Taylor expansion of e^{λ} is

$$e^{\lambda} = \sum_{x=0}^{\infty} \frac{\lambda^x}{x!}$$

So,
$$e^{\lambda}e^{-\lambda}=1$$

Taylor expansions are your friend... super under-rated.

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$$E(X) = \sum_{x=0}^{\infty} x \frac{\lambda^x e^{-\lambda}}{x!} = e^{-\lambda} \sum_{x=0}^{\infty} x \frac{\lambda^x}{x!}$$

Need to get rid of that x... Note the index of the sum changed

$$= e^{-\lambda} \lambda \sum_{x=1}^{\infty} \frac{\lambda^{x-1}}{(x-1)!}$$
$$= \lambda e^{-\lambda} e^{\lambda}$$

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$$Var(X) = \lambda$$

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$$Var(X) = \lambda$$

$$E(X^{2}) = \sum_{x=0}^{\infty} x^{2} \frac{\lambda^{x} e^{-\lambda}}{x!} = e^{-\lambda} \sum_{x=0}^{\infty} x \frac{\lambda^{x}}{x!}$$

Need to get rid of x^2 apply the factoring trick 2x now

$$=e^{-\lambda}\lambda\sum_{x=1}^{\infty}x\frac{\lambda^{x-1}}{(x-1)!}=$$

Repeat that trick to get

$$=e^{-\lambda}(\lambda+1)\lambda\sum_{x=2}^{\infty}\frac{\lambda^{x-2}}{(x-2)!}$$

$$Var(X) = E(X^2) - E(X)^2 = [\lambda + \lambda^2] - [\lambda]^2 = \lambda$$