$$M(t) = \mathbb{E}\left[e^{tx}\right] = \sum_{x=1}^{\infty} e^{tx} (1-p)^{x-1}$$

$$= p \sum_{\chi=0}^{\infty} \ell^{\chi+1} \chi^{\chi+1}$$

$$= pe^{t} \sum_{\chi=0}^{\infty} ((1-p)e^{t})^{\chi}$$

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$$= (1-p)e^{t}$$

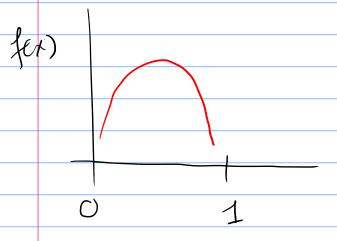
$$= pet\left(\frac{1}{1-(1-p)e^t}\right) |r| \times$$

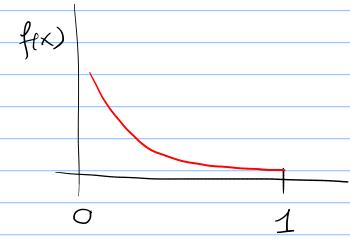
$$\mathbb{E}\left[\chi^2\right] = \frac{d^2M}{dt^2} = \frac{2-p}{p^2}$$

$$Var(X) = E[X^2] - E[X]^2 = \frac{2-p}{p^2} - (\frac{1}{p}) = \frac{1-p}{p^2}$$

Beta Distribution

- continuous dist w/ support on [0,1]





Beta Finction: B:R2 -> R

 $B(a,b) = \int_{0}^{1} \chi^{a-1} (1-\chi)^{b-1} d\chi$ 

$$=\frac{\Gamma(a)\Gamma(b)}{\Gamma(a+b)}$$

PDF! // Deta(a,b), a,b>0

$$f(x) = \frac{\chi^{a-1}(1-\chi)^{b-1}}{B(a-b)}$$

for 0 < X < 1

Expectation:

$$B(a+1,b) = \int_{0}^{1} \frac{x^{a-1}(1-x)^{b-1}}{B(a_1b)} dx$$

$$B(a+1,b) = \int_{0}^{1} \frac{x^{a}(1-x)^{b-1}}{B(a_1b)} dx$$

$$B(a+1,b) = \int_{0}^{1} \frac{x^{a}(1-x)^{b-1}}{B(a_1b)} dx$$

$$= \frac{B(a+1,b)}{B(a_1b)} = \frac{P(a+b+1)}{P(a+b+1)} P(a)P(b)$$

$$= \frac{a}{P(a+b)}$$

$$= \frac{a}{A+b} P(a+b)$$

$$= \frac{a}{A+b} P(a+b)$$

$$E[X'] = \int (X \times A^{-1}(1-X)^{b-1}) dX \qquad Arr-1 + b^{-1}$$

$$= \frac{B(a+r,b)}{B(a,b)} \left( \frac{X^{a+r-1}(1-x)^{b-1}}{X^{a+r-1}(1-x)^{b-1}} \right) dX$$

$$= \frac{B(a+r,b)}{B(a,b)} \left( \frac{X^{a+r-1}(1-x)^{b-1}}{X^{a+r-1}(1-x)^{b-1}} \right) dX$$

$$= \frac{B(a+r,b)}{B(a,b)}$$

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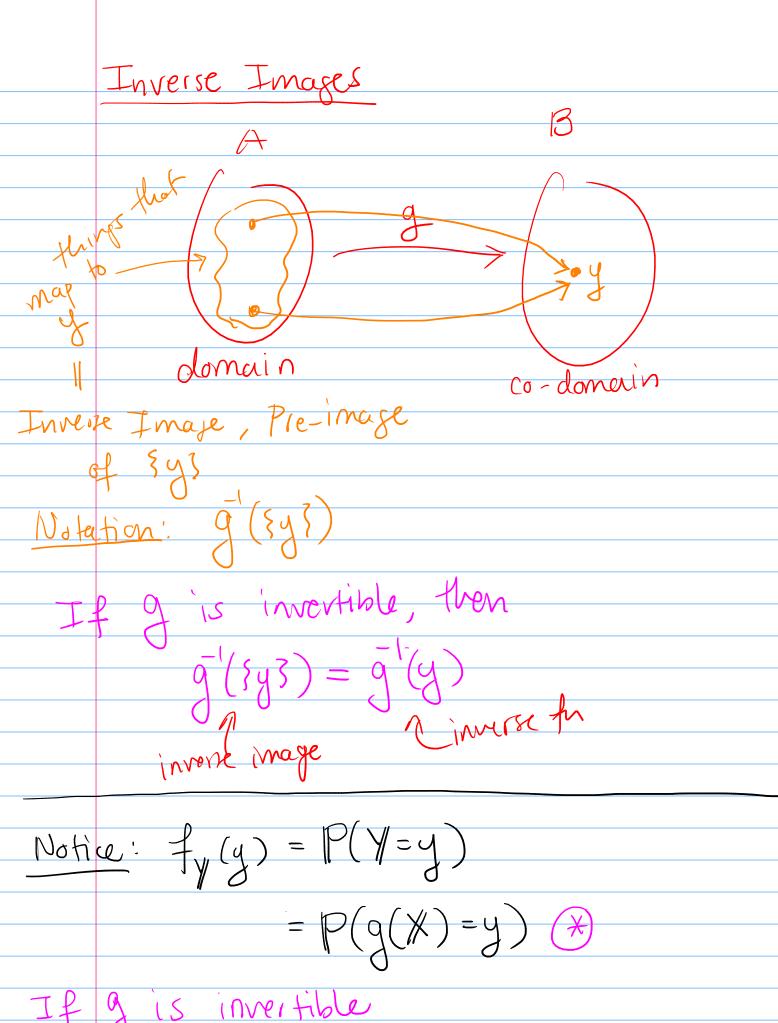
$$= \frac{B(a+2,b)}{B(a,b)}$$

$$= \frac{P(a+2)P(b)}{P(a+b)} \left( \frac{P(a)P(b)}{P(a+b)} \right) dX$$

$$= \frac{(a+1)aP(a)P(b)}{(a+b+1)(a+b)P(a+b)} \left( \frac{P(a)P(b)}{P(a+b)} \right) dX$$

$$= \frac{(a+1)aP(a)P(b)}{(a+b)(a+b+1)} \left( \frac{P(a)P(b)}{P(a+b)} \right) dX$$

$$= \frac{A(a+r)aP(a)P(b)}{(a+b)(a+b+1)} \left( \frac{P(a)P(b)}{P(a+b)} \right) dX$$



 $(x) = \mathbb{P}(x = g^{-1}(y))$ 

$$= \int_{X} (\bar{g}'(y)) P(X \in A)$$

$$= \sum_{x \in A} f(x)$$

$$= P(X \in \bar{g}'(y)) \text{ inverse image}$$

$$= \sum_{x \in \bar{g}'(y)} f_{x}(x)$$

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Theorem: If X is discrete and Y = g(X)

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$$f_{y}(y) = \int_{x:g(x)=y}^{x:g(x)=y} f_{x}(x)$$

$$= \sum_{x=n-y}^{x} f_{x}(x)$$

$$= f_{x}(n-y)$$

$$= (n-y) p^{x}(1-p)$$

$$= ($$

Continuous RVs: CDFs

Theorem: If X is cts and 
$$Y=g(X)$$

then

(1) If g is increasing then

 $F_{Y}(y) = F_{X}(g^{-1}(y))$ 

(2) If g is decreasing then

 $F_{Y}(y) = I - F_{X}(g^{-1}(y))$ .

Pf Case I: g is increasing

 $F_{Y}(y) = P(Y \leq y)$ 
 $= P(g(X) \leq y)$ 
 $= F_{X}(g^{-1}(y))$ 
 $= F_{X}(g^{-1}(y))$ 

Case 2: 
$$g$$
 dec.  

$$F_{y(y)} = P(Y \leq y)$$

$$= P(g(X) \leq y)$$

$$= P(X \geq g'(y))$$

$$= 1 - P(X \leq g'(y))$$

$$= 1 - F_{x}(g'(y))$$

$$= 1 - F_{x}(g'(y))$$

$$= 1 - F_{x}(g'(y))$$