Let Ll and L2 denote the first and second. litter.

Let B denote a brown havred vodent.

$$P(B|L1) = \frac{2}{3}$$
 and $P(B|L_2) = \frac{3}{5}$

Rondom sample from litter => P(Li) = P(Lz) = 1/2.

$$P(B) = P(L_1)P(B|L_1) + P(L_2)P(B|L_2)$$

= $\frac{1}{2} \times \frac{2}{3} + \frac{1}{2} \times \frac{3}{5}$
: $\frac{19}{10}$

ii) Use Bayls theorem:

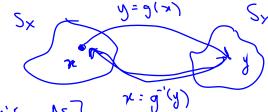
$$P(LI|B) = \frac{P(B|LI)P(U)}{P(B)}$$

$$= (\frac{2}{3}) \times (\frac{1}{2}) = \frac{10}{19} > 0.50$$



Proof: The transformation of 100 g(x) is monotone and each of goes to only one y and each y comes from out most one of (one-to-one from 5x ->5x)

Also, for each y tsy there is an XESX such that g(m) = y conto from Sx -> Sy)



The transformation g uniquely pairs n's and y's. As g is monotone than g^{-1} is singled valued that is $g^{-1}(y) = n$ iff y = g(x).

ci) If g is inveasing

$$\{x \in S_{\times} : g(x) \in y\} = \{x \in S_{\times} : g'(g(x)) \leq g^{-1}(y)\}$$

= $\{x \in S_{\times} : g'(y)\}$

Hence
$$F_{\gamma}(y) = \int_{\{x \in S_{\chi}: \chi \in g^{\dagger}(y)\}} f_{\chi}(x) dx$$

$$= \int_{-\infty}^{g^{-1}(y)} f_{\chi}(x) dx$$

$$\int_{a}^{b} f(n) dx = F(b) - F(a)$$

Sinct
$$F_{x}(-\infty) = 0$$
 $F_{x}(-\infty) = 0$



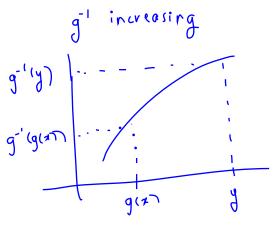
Case (ii) Now if g is monotonically decreasing then

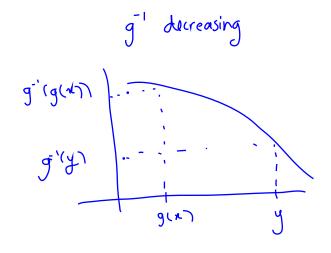
Fry = P(g(x) =y)

Y=g(X)

Notice that the inequality sign = has been flipped to >

Reason :





Hence when g is montonically decreasing and $g(x) \leq y$ by applying g^{-1} we see that $g^{-1}(g(x)) \geq g^{-1}(y)$.

Hence,
$$F_{y}(y) = \int_{\{x \in S_x: x \neq g'(y)\}} \int_{x \in S_x} \int_{x} (x) dx$$

$$= \int_{g'(y)}^{\infty} \int_{x \in S_x} \int_{x} (x) dx$$

$$= \lim_{x \to \infty} F_{x}(x) - F_{x}(g'(y))$$

$$= 1 - F_{\times}(g^{-1}(y))$$

O



Exercise 18

The joint density of X and Y

$$f_{XY}(x,y) = f_{X}(x)f_{Y}(y) = \frac{1}{\sqrt{2\pi}} e^{-x^{2}/2} \frac{1}{\sqrt{2\pi}} e^{-y^{2}/2}$$

independence

By solving U=n+y and V=n-y for x and y in terms of M and V:

$$M+V=M+y+(M-y)=2M=> M=h_1(M_1V)=\frac{M+V}{2}$$

 $M-V=(x+y)-(x-y)=2y=> y=h_2(M_1V)=\frac{M-V}{2}$

The Jacobian is given

$$J(M_1V) = \begin{vmatrix} \frac{\partial x}{\partial M} & \frac{\partial x}{\partial V} \\ \frac{\partial y}{\partial M} & \frac{\partial y}{\partial V} \end{vmatrix} = \begin{vmatrix} \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & -\frac{1}{2} \end{vmatrix} = (\frac{1}{2})(-\frac{1}{2}) - (\frac{1}{2})(\frac{1}{2}) = -\frac{1}{2}$$

The joint density of U and V

$$\int_{NV} (u_{1}v) = \int_{XY} \left(h_{1}(m_{1}v), h_{2}(m_{1}v) \right) \left| J(m_{1}v) \right| \\
= \frac{1}{2\pi T} \exp \left\{ -\frac{1}{2} \left(\frac{M+V}{2} \right)^{2} \right\} \exp \left\{ -\frac{1}{2} \left(\frac{M-V}{2} \right)^{2} \right\} \left| -\frac{1}{2} \right| -\infty < v < \infty$$

$$= \left(\frac{1}{42\pi T} \int_{Z} e^{2} \right) \left(\frac{1}{4\pi T} \int_{Z} e^{2} \right) \left(\frac{1}{4\pi T} \int_{Z} e^{2} \right) = \int_{M} (M) \int_{V} (v) dv = \int_{M} (M) \int_{W} (M) dv = \int_{M} (M) dv = \int_{M}$$

Statistical Inference MATH5905 T1 2021 – Chapter One

