

18) * p35 A necessary but not sufficient condition for independence is that the support is a cross product. Th 2.2a) If the support is not a cross product, then the RV's are dependent. The RV's could be dependent or ind if the support is a cross product.

19) Th 2.2 b) If y_1, \dots, y_n have cross product support, then they are ind iff $f(y_1, \dots, y_n) = h_1(y_1) \cdots h_n(y_n) \quad \forall y \in \mathcal{Y}$ where h_i is a positive function of y_i alone. In particular, if $\mathcal{Y} = \mathcal{Y}_1 \times \mathcal{Y}_2$, $y_1 \perp\!\!\!\perp y_2$ iff $f(y_1, y_2) = h_1(y_1) h_2(y_2) \quad \forall (y_1, y_2) \in \mathcal{Y}$ where $h_i(y_i) > 0$ for $y_i \in \mathcal{Y}_i$ and $i=1, 2$.

Th 2.2 c) y_1, \dots, y_n are ind iff $f(y_1, \dots, y_n) = g_1(y_1) \cdots g_n(y_n) \quad \forall y$ where $g_i(y_i) \geq 0$ is a function of y_i alone.

skip Proof a) If the support is not a cross product, $\exists y \exists f(y) = 0$ but $f_{y_i}(y_i) > 0$ for $i=1, \dots, n$. So $0 = f(y) \neq \prod_{i=1}^n f_{y_i}(y_i) > 0$.

so take $y \in \mathcal{Y} \setminus (\mathcal{Y}_1 \times \mathcal{Y}_2 \times \dots \times \mathcal{Y}_n)$

b) Proof for pdf, For pmf replace integrals by sums.

If y_1, \dots, y_n are ind, take $h_i(y_i) = f_{Y_i}(y_i) > 0$ for $y_i \in g_i$, $i=1, \dots, n$.

If $f(\underline{y}) = \prod_{i=1}^n h_i(y_i)$ for $\underline{y} \in \underline{g} = g_1 \times \dots \times g_n$, then $f(\underline{y}) = 0 = \prod_{i=1}^n f_{Y_i}(y_i)$ for $\underline{y} \notin \underline{g}$. So if

$$f(\underline{y}) = \prod_{i=1}^n f_{Y_i}(y_i) = \prod_{i=1}^n h_i(y_i) \text{ for } \underline{y} \in \underline{g}, \text{ we are done.}$$

$$\text{Now } 1 = \int_{\underline{g}} \int f(\underline{y}) d\underline{y} = \prod_{i=1}^n \underbrace{\int_{g_i} h_i(y_i) dy_i}_{a_i} = \prod_{i=1}^n a_i.$$

For $y_i \in g_i$, get marginal pdf by integrating out $y_1, \dots, y_{i-1}, y_{i+1}, \dots, y_n$. So

$$\begin{aligned} f_{Y_i}(y_i) &= \int_{g_n} \dots \int_{g_{i+1}} \int_{g_{i-1}} \dots \int_{g_1} \prod_{j=1}^n h_j(y_j) dy_1 \dots dy_{i-1} dy_{i+1} \dots dy_n \\ &= h_i(y_i) \prod_{\substack{j=1 \\ j \neq i}}^n \int_{g_j} h_j(y_j) dy_j = h_i(y_i) \prod_{\substack{j=1 \\ j \neq i}}^n a_j = h_i(y_i) \frac{1}{a_i}. \end{aligned}$$

Since $\prod_{j=1}^n a_j = 1$ and $a_i f_{Y_i}(y_i) = h_i(y_i)$ for $y_i \in g_i$,

$$\begin{aligned} f(\underline{y}) &= \prod_{i=1}^n h_i(y_i) = \prod_{i=1}^n a_i f_{Y_i}(y_i) = \left(\prod_{i=1}^n a_i \right) \prod_{i=1}^n f_{Y_i}(y_i) \\ &= \prod_{i=1}^n f_{Y_i}(y_i) \quad \text{if } \underline{y} \in \underline{g}. \end{aligned}$$

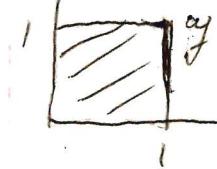
c) By a), the support is a cross product.

By b), take $g_i(y_i) = \begin{cases} h_i(y_i) & y_i \in g_i \\ 0 & \text{else} \end{cases}$.

20] c) is taught in M483, but factorizing on the support is simpler.

ex) $f(x, y) = 4xy$, $0 < x < 1, 0 < y < 1$

Support is a cross product



$$= 2x \cdot 2y = h_1(x)h_2(y) \text{ so } X \perp\!\!\!\perp Y.$$

$$= 4xy I[\{x \in (0,1)\} \cap \{y \in (0,1)\}]$$

$$= 4x I[x \in (0,1)] y I[y \in (0,1)]$$

$$g_1(x)$$

$$g_2(y)$$

$$\text{using } g = g_1 \times g_2.$$

Note $I[(x,y) \in g] = I[\{x \in g_1\} \cap \{y \in g_2\}] = I(x \in g_1) I(y \in g_2)$ since

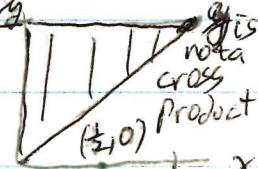
$$\text{RHS} = \text{LHS} = 1 \text{ iff } x \in g_1 \text{ and } y \in g_2 \text{ iff } (x,y) \in \{x \in g_1\} \cap \{y \in g_2\}.$$

Warning: Use indicators to denote support! the indicators must factor for a cross product.

ex) $f(x, y) = 8xy$

$$0 < x < y < 1$$

\Rightarrow not a cross product



so $f(x, y) = (8x)(y)$ for but X and Y are dependent.

$$I(0 < x < y < 1) \neq I(x \in g_1) I(y \in g_2)$$

ex) For HW2 # 6 $X \perp\!\!\!\perp Y$, $Z = X + Y$.

$$f_{X|Z=z}(x|z) = \frac{P(X=x, Z=z)}{P(Z=z)} = \frac{P(X=x, Y=z-x)}{P(Z=z)}$$

$X \perp\!\!\!\perp Y$

$$\stackrel{+}{=} P(X=x) P(Y=z-x)$$

$$P(Z=z)$$

2) Th 2.3 y_1, \dots, y_n ind \Rightarrow y_{i1}, \dots, y_{ik} are ind and

ii) $f(y_{i1}, \dots, \underbrace{y_{ik}}_{\text{no overlap}} | y_{j1}, \dots, y_{jm}) = f(y_{i1}, \dots, y_{ik})$.

22) * $E[h(\underline{y})] = \int_{\mathbb{R}^n} \dots \int h(\underline{y}) f(\underline{z}) d\underline{y}$ pdf

$$\sum_{i=1}^m \sum_{j=1}^{k_i} h(\underline{y}) f(\underline{z})$$

pmf

23) Th 2.4 $E[h(y_{i1}, \dots, y_{ik})] = \int_{\mathbb{R}^k} \dots \int h(y_{i1}, \dots, y_{ik}) f(y_{i1}, \dots, y_{ik}) dy_{i1} \dots dy_{ik}$

And also indicate how $\int_{\mathbb{R}^k} \dots \int$ for pmf

24) P38 Application: given $f(x, y)$ find
marginals $f_x(x)$ and $f_y(y)$, then 8.5

$$EY = \int y f_y(y) dy \quad \text{and} \quad EX = \int x f_x(x) dx.$$

25)* p39 If Y_1, \dots, Y_n are ind,

$$E[h_1(Y_1) \dots h_n(Y_n)] = \prod_{i=1}^n E[h_i(Y_i)] = E[h_1(Y_1)] \dots E[h_n(Y_n)]$$

proof LHS = $\int \prod h_i(y_i) f(y) dy = \prod \int h_i(y_i) f_i(y_i) dy_i = \prod E[h_i(Y_i)] = RHS$

$$\text{In particular, } Y_1 \perp\!\!\!\perp Y_2 \Rightarrow E(Y_1 Y_2) = E[Y_1] E[Y_2].$$

$$\begin{aligned} 26)* p40 \quad \text{cov}(Y_1, Y_2) &= E[(Y_1 - E(Y_1))(Y_2 - E(Y_2))] \\ &= E(Y_1 Y_2) - E(Y_1) E(Y_2). \end{aligned}$$

27) Th 2.9 If $\text{cov}(Y_1, Y_2)$ exists and $Y_1 \perp\!\!\!\perp Y_2$,
 $\text{cov}(Y_1, Y_2) = 0$, but $\text{cov}(Y_1, Y_2) = 0$ does
not mean that $Y_1 \perp\!\!\!\perp Y_2$.

- 28) common EI problem Given table for $f(y_1, y_2)$ pmf
or joint pdf $f(y_1, y_2)$
- IS $Y_1 \perp\!\!\!\perp Y_2$?
 - Find marginals f_{y_1}, f_{y_2}
 - Find $EY_1, EY_2, EY_1 Y_2, \text{cov}(Y_1, Y_2)$.

see ex 2.6, 2.7, 2.8.

29)* Th 2.7 If Y_1, \dots, Y_n are ind
then $h_1(Y_1), \dots, h_n(Y_n)$ are ind
crossproduct

$$\begin{aligned} \text{ex)} \quad \text{If } X \perp\!\!\!\perp Y, \quad P(X \in A, Y \in B) &= E[I_{\{(X,Y) \mid X \in A, Y \in B\}}] \\ &= E[I_A(X) I_B(Y)] = E[I_A(X)] E[I_B(Y)] = P(X \in A) P(Y \in B) \end{aligned}$$

30) Know $I_A(x) \sim \text{Bernoulli}$ with $P(I_A(x)=1) = P(X \in A)$.

§2.3 3) p41 $Y|X=x$ is a single distribution, 9

$Y|X$ is a family of distributions.

ex) $Y|X=x \sim N(c+dx, \sigma^2)$, $Y|X \sim N(c+dX, \sigma^2)$
is the family of normal dist's with variance σ^2
and mean $\mu_{Y|X=x} = c+dx$.

32) p42 $W \equiv w_x \equiv Y|X=x$ has pmf $f(y|x) \equiv f_w(y) \equiv f_{w_x}(y)$

The conditional expected value

$$E[h(Y)|X=x] = \begin{cases} \sum_y h(y) f(y|x) & \text{pmf} \\ \int_{-\infty}^{\infty} h(y) f(y|x) dy & \text{pdf} \end{cases}$$

$$\text{In particular } E(Y|X=x) = \begin{cases} \sum_y y f(y|x) & \text{pmf} \\ \int_{-\infty}^{\infty} y f(y|x) dy & \text{pdf} \end{cases}$$

33) p42 The conditional variance

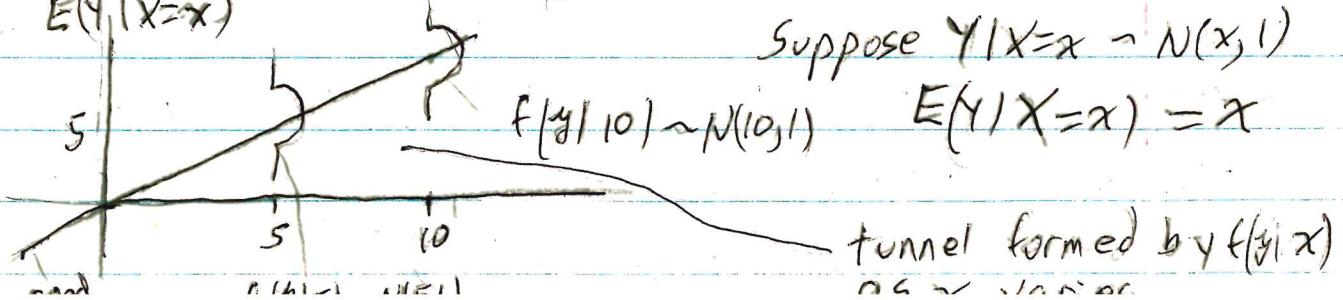
$$\text{Var}(Y|X=x) = E(Y^2|X=x) - [E(Y|X=x)]^2$$
$$= E[(Y - E(Y|X=x))^2 | X=x]$$

Note $V(Y|X=x) = E(w_x^2) - [E(w_x)]^2$

34) p42 $f(y|x)$ is a function of y with x fixed,
but $E(Y|X=x) = m(x)$ is a function of x .
 $E(Y|X) = m(X)$ and $V(Y|X) = v(X)$ are
random variables.

ex) $Y = \text{weight}$ $X = \text{age}$ ^{of people in Illinois} $E(Y|X=x)$ is
the expected weight of all people of age x
in Illinois. $E(Y|X=4) < E(Y|X=18)$.

ex) $E(Y|X=x)$ Suppose $Y|X=x \sim N(x, 1)$



ex] If $P(X=5) = P(X=10) = \frac{1}{2}$ in last ex) then^{a.5}
 perform the experiment resulting in $X=x$.
 Sometimes $x=5$ and sometimes $x=10$.

35] Know p 43 Iterated Expectations

$$E(Y) = E[E(Y|X)]$$

(or $E_Y(Y) = E_X[E_{Y|X}(Y|X)]$) if the expectations exist.

$m(x)$ where $m(x) = E(Y|X=x)$

Proof for pdf: $E(Y) = \iint y f(x,y) dy dx$

$$\begin{aligned} &= \iint y f(y|x) f(x) dy dx = \int \left[\int y f_{Y|X}(y|x) dy \right] f_X(x) dx \\ &= \int \underbrace{E(Y|X=x)}_{m(x)} f_X(x) dx = E[m(x)] \\ &= E[E(Y|X)]. \quad \square \end{aligned}$$

36] Know P43 Steiner's Formula = Conditional Variance

Identity: $V(Y) = E[V(Y|X)] + V[E(Y|X)]$

provided the expected values exist.

Read proof on p 51 - 2.

37] E1 problem p 43 given $Y|X \sim W$
 and $X \sim U$, find $E(Y)$ and $V(Y)$.

ex] See ex 2.10, 2.68 for a qual problem

ex] 2.73 $Y|P=p \sim \text{binomial}(k,p)$, $P \sim \text{beta}(\delta=4, v=6)$
 Find a) $E(Y)$, b) $V(Y)$