Independent variables

Consider random variable X, and random variable y, the probability that X is between x+dx and y is between y+dy is

JP = P(x, y) dx dy

If x and y are independent then the P(x, 4) factorizes

dP = Px(x) dx Py(y)dy

Then

<xy> = \[dxdy P(x,y) xy \]

= \int dx dy Px (x) Py (4) xy

= Jdx Px(x) x Jdy Py(4)dy y

= < x>< y>

Sums of independent variables

· Consider

Y = x, + x₂ + x_n

where each x is drawn independently from the distribution P(x) i.e $P(x_1, \dots, x_n) = P(x_1) P(x_2) \dots P(x_n) dx_1 dx_2 \dots dx_n$ Then Clear n times the average of (x) < >> = < x, > + < x, > + ... < x, > = n < x > What about the Variance of XY>? · SY = Y - <Y> $= (\times, -\langle \times \rangle) + (\times, -\langle \times \rangle) + \dots \times (\times, -\langle \times \rangle)$ $\delta Y = \delta X + \delta X + \dots + \delta X$ (8Y2) = < (8x, +8x, + ... 8x,)) = $\langle 8x_i^2 \rangle + \langle 8x_i^2 \rangle + ... \langle 8x_n^2 \rangle + terms | ike$ < 5×, 8×2) But the cross terms all vanish since (Sx, 8x2) = (8x,) (8x2) = 0 independence

So

02 = (8/2) = n < x2) = n 0x

There is more that can be said. I will not prove it, but if n is large the probability of Y is gaussian, regardless of P(x)!

This is the Central Limit Theorem

 $P(Y) = \frac{1}{\sqrt{2\pi\sigma_{Y}^{2}}} = \frac{(Y - \overline{Y})^{2}/2\sigma_{Y}^{2}}{\sqrt{2\pi\sigma_{Y}^{2}}}$ with $\sigma_{Y}^{2} = n \sigma_{X}^{2}$ $Y = n\langle x \rangle$