Probability generating function (.)
Fundamental question: X,, X2,, Xn independent random
Variables.
Variables. What is the distribution of $\sum_{i=1}^{n} \chi_{i}^{2}$
For n=2 can use completion formula, then iteratively extend to n72. In practice, not efficient, teclions, and for large n too time consuming. Better tools are the following three transforms:
1.) Probability generaling function (good for integer-value random variables)
2.) Moment generaling function, mgf
3.) Characteristic function.

Det: Let X be a non-negative, integer-valuel?

Yandon variable. The probability generally
function of X is defined as

$$\begin{aligned}
& \int_{X_{-}}^{X_{-}} (6) \circ = \mathbb{E}[\mathcal{L}^{X}] \\
& = \sum_{n=0}^{\infty} \mathcal{L}^{n} \cdot \mathbb{P}(X=n) \cdot \mathcal{L}^{n} \\
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Remark: At least for 16/51, the sum is absolutely

Convergent, because

$$\sum_{n=0}^{\infty} |\mathcal{Z}^n| \cdot \mathbb{P}(X=n) \leqslant \sum_{n=0}^{\infty} \mathbb{P}(X=n) = \mathcal{J}.$$

In particular: $g_X(1) = 1$.

Ex: $X \in Be(p)$, P(X=0) = q, P(X=1) = p, p+q=1. $g_{X}(t) = f[t^{x}] = f'q + f'p = q + fp$.

Than 1: Let X and Y be non-negative integer-valued random variables.

If $g_{\chi} = g_{\chi}$, then $Y = \chi$

This 2: Let X1, X2, --, Xn be independent, non-negative integ-valued random variables. Set

Sn:= X1+X2+ ...+ Xn.

Then $g_{S_n}(b) = 11 g_{X_k}(b)$ $k=1 g_{X_k}(b)$

Proof! Observe that $\pm x_1 \pm x_2 + \dots + x_n = \text{The are independent}$ $\int_{Sn}^{\infty} (t) = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots + x_n \right] = \int_{\mathbb{R}^n} \left[\pm x_1 + x_2 + \dots$

$$0 \times 1 + \times 2^{(6)} = 9 \times (6) - 9 \times (6) = (9 + 6)^{2}.$$

Ptg=1.

$$= \sum_{k=0}^{n} \binom{n}{k} (\xi p)^{k} q^{n-k}$$

Binomial identity:
$$(a+b)^n = \sum_{k=0}^{h} \binom{n}{k} a^k b^{n-k}$$
, $a_1b \in \mathbb{R}$, $n \in \mathbb{N}$.

$$\binom{n}{k} = \frac{n!}{k! (n-k)!}$$

$$\Im \chi^{(6)} = \sum_{n=0}^{\infty} \ell^n P(\chi=n)$$

terr

(5.)

Take derivative int

 $P(X=0) + EP(X=1) + E^2P(X=2) + ----$ (ok, for 1614)

$$Q(x) = \frac{d}{dt} Q(x) = \sum_{n=1}^{(6k, f_n)} \frac{(6k^2)}{n!} = \sum_{n=$$

 $= P(x=1) + 2 L P(X=2) + 36^2 P(x=3) + \dots$

$$\int_{0}^{\infty} \chi''(4) = \int_{0}^{\infty} n(h-1) + \int_{0}^{\infty} P(\chi=3) + 4.3 + \int_{0}^{\infty} P(\chi=4) + \dots$$
Chows now $f = 0$.

$$g_{\chi}(0) = P(\chi=1), g_{\chi}(0) = 2.1.P(\chi=2)$$

$$P(X=n) = \underbrace{9_X^{(n)}(0)}_{N!} = \underbrace{\frac{1}{n!}}_{N!} \underbrace{\frac{1}{dt^n}}_{M!} \underbrace{9_X^{(6)}}_{t=0}$$
we now Theorem 1 ensing this regular

Exercise: Prove now Theorem I ensing this result.

$$g_{\chi}(1) = \sum_{n=1}^{\infty} n \mathbb{P}(\chi=n) = \mathbb{F}[\chi]$$

$$F[X] = g[X(1)]$$

Similarly,

$$Var(X) = F(X - FX)^2 = g_X^{(1)}(1) + g_X^{(2)}$$

(provided that mean and second moment exist). - (9x(1))2

Ex: KEBin(n,p)

$$H(X) = g_X'(1) = d_t (q+pt)^{n-1} |_{t=1}$$
= $n (q+pt)p|_{t=1}$
= $n \cdot p \cdot q+p=1$

$$g_{\chi}(6) = \sum_{k=0}^{\infty} \xi_{k} e^{k} e^{k} \frac{m_{k}}{k!}$$

$$= e^{m} \sum_{k=0}^{\infty} (\xi_{m})_{k}$$

Exercise: Compute IX, VarX.

Find the distribution of $X_1 + X_2$ when $X_1 \in P_0(m_1)$ and $X_1 \in P_0(m_2)$, and expendent.

Answers: EX = m, Var X = m $X_1 + X_2 \in Po(m_1 + m_2)$