Remark: under the same setup i.e. XI, XI, un correlated with EX:= Wi and v(xi) = Ti suppose $\Sigma T_i^* \rightarrow V$, then we can take $a_n = \Sigma T_i^*$ $b_n = \Sigma T_i^*$ $\frac{S_n - a_n}{b_n} \stackrel{p}{\longrightarrow} 0$ Remark: If Ti= T2 +i, then the condition $\frac{1}{n^2} \sum \overline{\tau}_i^2 \rightarrow 0 \text{ as } n \rightarrow 4 \text{ is automatically}$ satisfies and WLLN holds Remark: of X1, -- . in a seq of i.i.d. v.v. s with mean u and variance TIZH, then WLLN holds for {Xn} + Xx PX = EX Note that firsteners of varionce is not read by 'Khintchine's WLLN.

Convergence in distribution (or law) Del": He say that a sea of r-r-s {xn} converges in distribution to X (Xn xx x x x x) Yn at which the if $F_{\chi_n}(x) \longrightarrow F(x)$ limiting dist t is untinums: $F_{x_n}(\cdot)$: $d \cdot f \cdot f \times n$ $F_{x_n}(\cdot)$: $d \cdot f \cdot f \times n$ Examples
(1) 11 X1, - -1 Xn 1.1.d. N(0,1) $\lambda^{\nu} = \frac{1}{\nu} \sum_{i=1}^{\infty} x_i = x^{\nu}$ $\lambda^{\nu} = \frac{1}{\nu} \sum_{i=1}^{\infty} x_i = x^{\nu}$ $\mathsf{E}^{\lambda^{\nu}}(\mathsf{A}) = \mathsf{b}(\lambda^{\nu} \in \mathsf{A}) = \mathsf{b}(\Lambda^{\nu} \lambda^{\nu} \in \mathsf{L}^{\nu} \mathsf{A})$ $= \Phi(1 \times 3) \rightarrow \begin{cases} 0 & 3 < 0 \\ \frac{1}{2}, & 3 > 0 \end{cases}$ Consider a d.t Fy (a) = \ 0 \ A > 0 i.e. P(Y=0)=1

From

$$F_{y_n}(y) \rightarrow F_{y_n}(y). \quad \forall y \text{ at which } F_{y_n}(.) \text{ is } Continuous}$$

$$\Rightarrow y_n \xrightarrow{\lambda} y \leftarrow a \text{ degenerate } \tau.v.)$$

$$i.e. \quad y_n \xrightarrow{\lambda} 0$$

$$2) \quad x_{i,x_2}... \quad i.i.d. \quad U(0,0); \quad \theta > 0$$

$$y_n = x_{(n)} = \max\{x_{i,x_2}...,x_n\}; \quad y_n \xrightarrow{p} \theta$$

$$F_{y_n}(y) = \begin{cases} 0, & y < 0 \\ \frac{y}{\theta} \end{pmatrix}, & 0 \le y \le \theta \end{cases} \Rightarrow \begin{cases} 0, & y < \theta \\ 1, & y > \theta \end{cases}$$

$$\Rightarrow y_n \xrightarrow{\lambda} y, \text{ a degenerate } \tau.v.; \text{ degenerate at } \theta$$

$$\lim_{x \to \infty} \frac{x_{i,x_2}... \cdot x_{i,x_3}}{x_{i,x_3}} = n(\theta - y_n)$$

$$F_{z_n}(x) = P(z_n \le x)$$

$$= P(n(\theta - x_{(n)}) \le x)$$

$$= P(x_{(n)} \ge \theta - \frac{x}{n}) = 1 - F_{x_{(n)}}(x_n - \theta - \frac{x}{n})$$

 $= \begin{cases} 0, & x < 0 \\ 1 - \left(\frac{\theta - x/n}{\theta}\right)^n & 0 \le x \le n\theta \end{cases}$

$$F_{Z_n}(x) \rightarrow \begin{cases} 0, & x < 0 \\ 1 - e^{-x/\theta}, & x > 0 \end{cases}$$

$$i.e. \quad Z_n \rightarrow Z \quad \text{where } f(3) = \begin{cases} 0, & 3 < 0 \\ 1 - e^{-3/\theta}, & 3 > 0 \end{cases}$$

$$i.e. \quad Z_n \rightarrow Z \sim \exp(0, \theta)$$
Scale θ

(3)
$$\{x_n\} - x_{eq} \neq discrete r.v.s.$$

$$P(x_n = x) = \{1, \frac{1}{2} \neq x = 2 + \frac{1}{2} \}$$

$$F_{x_n}(x) = \{0, \frac{1}{2} \neq x = 2 + \frac{1}{2} \}$$

$$x < 2 + \frac{1}{2} \}$$

$$x > 2 + \frac{1}{2} \}$$

convergence in law can also be proved using m.q.f.

(4)
$$X_n \sim Bin(n, \theta)$$

Suppose $n \rightarrow \alpha \rightarrow n \beta = \lambda$ in fixed i.e. $\theta = \frac{\lambda}{n}$
 $M_{X_n}(E) = ((1-\theta) + \theta e^E)^n$

$$= (1 + \frac{\lambda}{n}(e^{t}-1))^{n}$$

$$\rightarrow e^{\theta(e^{t}-1)} \Leftrightarrow n \Rightarrow \alpha$$

(5)
$$X_1, \dots, X_n \sim N(0, \frac{1}{n})$$

$$\overline{X}_n \sim N(0, \frac{1}{n})$$

$$M_{\overline{X}_n}(t) = e^{\frac{t^2}{2n}} \rightarrow 1 \text{ as } n \rightarrow t$$

=> Xn / X; where x in degenerate at 0

> Xn -> X; Where X~P(A)