

# THE BREIMAN CONJECTURE

DAVID M. MASON \*

UNIVERSITY OF DELAWARE

\*Most of the results in this talk are based on joint work with Péter Kevei and Joel Zinn.

# RANDOMLY WEIGHTED SELF-NORMALIZED SUMS

Let  $\{Y, Y_i : i \geq 1\}$  denote a sequence of i.i.d. random variables, where  $Y$  is non-negative with cumulative distribution function [c.d.f.]  $G$ .

Now let  $\{X, X_i : i \geq 1\}$  be a sequence of i.i.d. random variables, independent of  $\{Y, Y_i : i \geq 1\}$ , where  $X$  is in the class  $\mathcal{X}$  of non-degenerate random variables satisfying for  $X \in \mathcal{X}$

$$\mathbb{E}|X| < \infty.$$

For future use, let  $\mathcal{X}_0$  denote those  $X \in \mathcal{X}$  such that  $\mathbb{E}X = 0$ . Consider the randomly weighted self-normalized sum

$$\mathbb{T}_n = \sum_{i=1}^n X_i Y_i / \sum_{i=1}^n Y_i.$$

# RANDOMLY SIGNED SELF-NORMALIZED SUMS

Here is a motivating special case.

Let  $\{Y, Y_i : i \geq 1\}$  and  $\{s, s_i : i \geq 1\}$  be independent sequences of random variables, where the  $Y_i$ 's are i.i.d.  $Y$  positive and the  $s_i$ 's are i.i.d.  $s$ , where  $s$  is the random sign

$$P\{s = 1\} = P\{s = -1\} = 1/2.$$

Consider the randomly signed self-normalized sum,

$$\mathbb{T}_n := \sum_{i=1}^n s_i Y_i / \sum_{i=1}^n Y_i.$$

# THE ARCSINE LAW

The randomly signed self-normalized sum has this interesting motivation. In fair coin tossing, where  $+1$  denotes heads and  $-1$  tails, let  $Y_i$  be the time between the  $(i - 1)^{th}$  and  $i^{th}$  return to zero of the partial sums  $S_1, S_2, \dots$ , of the coin toss outcomes. Then

$$(\mathbb{T}_n + 1) / 2$$

is the fraction of the time at the  $n^{th}$  return to zero that the sums were positive.

In this setup  $(\mathbb{T}_n + 1) / 2$  asymptotically has the arcsine law, namely for all  $0 \leq x \leq 1$ ,

$$P \{ (\mathbb{T}_n + 1) / 2 \leq x \} \rightarrow \frac{2}{\pi} \arcsin (\sqrt{x}) .$$

# DOMAIN OF ATTRACTION

In this talk

$$Y \geq 0 \text{ and } Y \in D(\beta), 0 \leq \beta < 1,$$

means that for some function  $L$  slowly varying at infinity,

$$\overline{G}(y) = y^{-\beta} L(y), \quad y > 0,$$

where for any c.d.f.  $G$

$$\overline{G}(y) := P\{Y > y\}.$$

In the case  $0 < \beta < 1$  this is equivalent to  $Y$  being in the domain of attraction of a stable law of index  $\beta$ .

## BREIMAN RESULT

Among other results, Breiman (1965) proved that  $\mathbb{T}_n$  converges in distribution for EVERY  $X \in \mathcal{X}$  with at least one limit law being non-degenerate if and only if

$$Y \in D(\beta), \text{ with } 0 \leq \beta < 1. \quad (1)$$

## BREIMAN CONJECTURE [BC]

At the end of his paper Breiman conjectured that if for *some*  $X \in \mathcal{X}$ ,  $\mathbb{T}_n$  converges in distribution to some nondegenerate random variable  $T$ , written

$$\mathbb{T}_n \rightarrow_d T, \text{ as } n \rightarrow \infty, \text{ with } T \text{ nondegenerate,} \quad (2)$$

then (1) holds.

## OBSERVATION

By Proposition 2 and Theorem 3 of Breiman (1965), for any  $X \in \mathcal{X}$ , (1) implies (2), in which case  $T$  has a distribution related to the arcsine law. Using this fact, we see that his conjecture can restated to be: for any  $X \in \mathcal{X}$ ,

(1) is equivalent to (2).



## A PARTIAL SOLUTION

It has proved to be surprisingly challenging to resolve. Using Karamata's Tauberian theorem, M and Zinn [MZ] (2005) partially verified the Breiman conjecture.

They established that whenever  $X$  is nondegenerate and satisfies  $\mathbb{E}|X|^p < \infty$  for some  $p > 2$ , then (1) is equivalent to (2). The  $p > 2$  moment condition was imposed in order to conclude that

$$\mathbb{E} \left( \mathbb{T}_n^2 \right) \rightarrow \mathbb{E} \left( T^2 \right) < \infty, \text{ as } n \rightarrow \infty.$$

## SLIGHT EXTENSION OF MZ

Here is a slight extension of their proof, showing that  $\mathbb{E}|X|^2 < \infty$  suffices. Without loss of generality we can assume that  $\mathbb{E}X = 0$ . An easy calculation gives

$$\mathbb{E}(\mathbb{T}_n)^2 = \text{var}(X) n \mathbb{E} \left( \frac{Y_1}{Y_1 + \cdots + Y_n} \right)^2.$$

## LEMMA

**Lemma** *Assume that*

$$\mathbb{T}_n \rightarrow_d T, \text{ as } n \rightarrow \infty,$$

*where  $T$  is random variable. Whenever for some  $p \geq 1$ ,  $\mathbb{E} |X|^p < \infty$ , then*

$$\mathbb{E} |\mathbb{T}_n|^p \rightarrow \mathbb{E} |T|^p < \infty, \text{ as } n \rightarrow \infty.$$

## IN PARTICULAR

In particular, when  $\mathbb{E}|X|^2 < \infty$

$$\mathbb{E} \left( \mathbb{T}_n^2 \right) \rightarrow \mathbb{E} \left( T^2 \right) < \infty, \text{ as } n \rightarrow \infty,$$

and thus whenever  $\mathbb{E}X = 0$  and  $T$  is nondegenerate, for some  $0 \leq \beta < 1$ ,

$$\begin{aligned} \mathbb{E} \left( \mathbb{T}_n^2 \right) &= \text{var} (X) n \mathbb{E} \left( \frac{Y_1}{Y_1 + \cdots + Y_n} \right)^2 \\ &\rightarrow \text{var} (X) (1 - \beta). \end{aligned}$$

## MOMENT RESULT

Arguing as in MZ we get that  $\mathbb{E}|X|^2 < \infty$  and (2) suffice for (1), using the following moment result due to Fuks, Ioffe and Teugels (2001), which is proved using Tauberian theorems.

**Proposition** We have  $Y \in D(\beta)$ , with  $0 \leq \beta < 1$ , if and only if

$$n\mathbb{E} \left( \frac{Y_1}{Y_1 + \cdots + Y_n} \right)^2 \rightarrow 1 - \beta.$$

## CRUCIAL TO THE PROOF

Crucial to the proof of this result was the representation

$$\begin{aligned} & n\mathbb{E} \left( \frac{Y_1}{Y_1 + \cdots + Y_n} \right)^2 \\ &= n \int_0^\infty u \varphi''(u) (\varphi(u))^{n-1} \mathrm{d}u, \end{aligned}$$

where  $\varphi(u) = \mathbb{E} \exp(-uY_1)$ , for  $u \geq 0$ .

## KEVEI-M RESULT

Kevei and M (2015) have further extended the MZ partial solution to the BC. In the following  $\phi_X(t)$  denotes the characteristic function of  $X$ .

**Theorem** *Assume that for some  $X \in \mathcal{X}_0$ ,  $1 < \alpha \leq 2$ , positive slowly varying function  $L$  at zero and  $c > 0$ ,*

$$\frac{-\log(\Re \phi_X(t))}{|t|^\alpha L(|t|)} \rightarrow c, \text{ as } t \rightarrow 0. \quad (3)$$

*Whenever (2) holds then  $Y \in D(\beta)$  for some  $\beta \in [0, 1)$ .*

## COROLLARY

Let  $\mathcal{F}$  denote the class of random variables that satisfy the conditions of the theorem. Applying our theorem in combination with Proposition 2 and Theorem 3 of Breiman (1965) we get the following corollary.

**Corollary** *Whenever  $X - \mathbb{E}X \in \mathcal{F}$ , (1) is equivalent to (2).*



## IMPORTANT OBSERVATIONS

It can be inferred from Theorem 8.1.10 of Bingham, Goldie, and Teugels (1987) that for  $X \in \mathcal{X}_0$ , (3) holds for some  $1 < \alpha < 2$ , positive slowly varying function  $L$  at zero and  $c > 0$  if and only if  $X$  satisfies

$$\mathbb{P}\{|X| > x\} \sim L(1/x)x^{-\alpha}c\Gamma(\alpha)\frac{2}{\pi}\sin\left(\frac{\pi\alpha}{2}\right).$$

Note that a random variable  $X \in \mathcal{X}_0$  in the domain of attraction of a stable law of index  $1 < \alpha < 2$  satisfies (3).

Also a random variable  $X \in \mathcal{X}_0$  with variance  $0 < \sigma^2 < \infty$  fulfills (3) with  $\alpha = 2$ ,  $L = 1$  and  $c = \sigma^2/2$ . This means that the Kevei-M theorem contains the MZ result.

## PROPOSITION 1

The theorem is a consequence of the two propositions that follow. First we need more notation. For any  $\alpha \in (1, 2]$  define for  $n \geq 1$

$$S_n(\alpha) = \frac{\sum_{i=1}^n Y_i^\alpha}{(\sum_{i=1}^n Y_i)^\alpha}.$$

**Proposition 1** *Assume that the assumptions of the theorem hold. Then for some  $0 < \gamma \leq 1$*

$$\mathbb{E}S_n(\alpha) \rightarrow \gamma, \text{ as } n \rightarrow \infty. \quad (4)$$

## PROPOSITION 2

The next proposition is interesting in its own right. It is an extension of Theorem 5.3 of Fuchs, Joffe and Teugels (2002), where  $\alpha = 2$  (see also Proposition 3 of MZ).

**Proposition 2** *If (4) holds with some  $\gamma \in (0, 1]$  then  $Y \in D(\beta)$ , for some  $\beta \in [0, 1)$ , where  $-\beta \in (-1, 0]$  is the unique solution of*

$$\frac{\Gamma(\alpha - 1)\Gamma(1 - \beta)}{\Gamma(\alpha - \beta)} = \frac{1}{\gamma(\alpha - 1)}.$$

*In particular,  $Y \in D(0)$  for  $\gamma = 1$ .*

*Conversely, if  $G \in D(\beta)$ ,  $0 \leq \beta < 1$ , then (4) holds with*

$$\gamma = \frac{\Gamma(\alpha - \beta)}{\Gamma(\alpha)\Gamma(1 - \beta)}.$$

## KEY RESULT USED IN PROOF

Set for each  $n \geq 1$ , for  $i = 1, \dots, n$

$$R_{i,n} = Y_i / \sum_{l=1}^n Y_l.$$

Consider the sequence of strictly decreasing continuous functions  $\{\varphi_n\}_{n \geq 1}$  on  $[1, \infty)$  defined for  $y \in [1, \infty)$  by

$$\varphi_n(y) = \mathbb{E} \left( \sum_{i=1}^n R_{i,n}^y \right).$$

## LEMMA

Note that each function  $\varphi_n$  satisfies  $\varphi_n(1) = 1$ . By a diagonal selection procedure for each subsequence of  $\{n\}_{n \geq 1}$  there is a further subsequence  $\{n_k\}_{k \geq 1}$  and a right continuous non-increasing function  $\psi$  such that  $\varphi_{n_k}$  converges to  $\psi$  at each continuity point of  $\psi$ .

**Lemma** *Each such function  $\psi$  is continuous on  $(1, \infty)$ .*

## KEY STEP USED IN PROOF

One can show that

$$\mathbb{E} \frac{\sum_{i=1}^n Y_i^\alpha}{(\sum_{i=1}^n Y_i)^\alpha} = \frac{n}{\Gamma(\alpha)} \int_0^\infty u^{\alpha-1} \mathbb{E} \left( e^{-uY_1} Y_1^\alpha \right) (\mathbb{E} e^{-uY_1})^{n-1} du,$$

which if it converges to  $\gamma \in (0, 1]$ , then

$$t^{\alpha-1} \frac{\int_0^\infty \overline{G}(u) u^{\alpha-1} e^{-ut} du}{\Gamma(\alpha) \int_0^\infty \overline{G}(u) e^{-ut} du} \rightarrow \gamma, \text{ as } t \searrow 0. \quad (5)$$

# APPLICATION OF DRASIN-SHEA THEOREM

An application of a version of the Drasin-Shea theorem shows that (5) implies  $G \in D(\beta)$ , where  $0 \leq \beta < 1$  satisfies

$$\gamma = \frac{\Gamma(\alpha - \beta)}{\Gamma(\alpha)\Gamma(1 - \beta)}.$$

# MELLIN TRANSFORM -CONVOLUTION

**Mellin transform** Given a measurable kernel  $k : (0, \infty) \rightarrow \mathbb{R}$ , its Mellin transform is

$$\tilde{k}(z) = \int_{(0, \infty)} t^{-z-1} k(t) dt$$

for  $z \in \mathbb{C}$ , such that  $\tilde{k}(z)$  is finite.

**Mellin convolution** For suitable functions  $f$  and  $g : (0, \infty) \rightarrow \mathbb{R}$ ,

$$f \stackrel{M}{*} g(x) = \int_{(0, \infty)} t^{-1} f(x/t) g(t) dt.$$



## BOUNDED DECREASE

The **lower Matuszewska index**  $\beta(f)$  is the supremum of those  $\beta$  for which there exists a constant  $d = d(\beta) > 0$  such that for each  $\Lambda > 1$ , as  $x \rightarrow \infty$ , uniformly in  $\lambda \in [1, \Lambda]$ ,

$$f(\lambda x) / f(x) \geq d \{1 + o(1)\} \lambda^\beta.$$

**Bounded decrease [BD]** A positive function  $f$  has bounded decrease if  $\beta(f) > -\infty$ .

# DRASHIN-SHEA THEOREM

**Theorem** Let  $k$  be a non-negative kernel whose Mellin transform  $\tilde{k}$  has maximal convergence strip  $a < \Re z < b$ , where  $a < 0$ . If  $a$  is finite, assume that  $\tilde{k}(a+) = \infty$ , if  $b$  is finite assume  $\tilde{k}(b-) = \infty$ . Let  $f$  be a non-negative, locally bounded on  $[0, \infty)$ . Assume that  $f \in BD$ . If

$$k \overset{M}{*} f(x) / f(x) \rightarrow c, \text{ as } x \rightarrow \infty,$$

holds, then  $c = \tilde{k}(\rho)$  for some  $\rho \in (a, b)$ , and  $f \in R_\rho$ , meaning that  $f$  is regularly varying at infinity with index  $\rho$ .

## OUR KERNEL

The particular kernel that we use is

$$k(u) = \begin{cases} (u-1)^{\alpha-2} u^{-\alpha+1}, & u > 1, \\ 0, & 0 < u \leq 1. \end{cases}$$

Its *Mellin-transform* is

$$\begin{aligned} \tilde{k}(z) &= \int_1^\infty (u-1)^{\alpha-2} u^{-\alpha-z} du \\ &= \frac{\Gamma(\alpha-1) \Gamma(1+z)}{\Gamma(\alpha-z)}, \end{aligned}$$

which is convergent for  $z > -1$ .

## FURTHER RESULTS

Kevei and M (2012) studied the asymptotic distribution of randomly weighted sums  $\mathbb{T}_n$  along subsequences  $\{n'\}$  of  $\{n\}$ .

Our basic result is the following extension of Proposition 3 of Breiman (1965), who proved it in the case  $Y \in D(\beta)$ , with  $0 < \beta < 1$ , and for the full sequence  $\{n\}$ .

Assume that  $\mathbb{E}|X| < \infty$  and for a subsequence  $\{n'\}$  there exists a sequence of positive norming constants  $a_{n'}$  such that

$$\frac{1}{a_{n'}} \sum_{i=1}^{n'} Y_i \xrightarrow{D} W_2, \text{ as } n' \rightarrow \infty,$$

where  $W_2$  is an  $id(0, b, \Lambda)$  infinitely divisible random variable.

## $id(0, b, \Lambda)$ **RANDOM VARIABLE**

This means that  $W_2$  is an infinitely divisible random variable taking values in  $[0, \infty)$  with characteristic exponent

$$\begin{aligned} & \log \mathbb{E} e^{iuW_2} \\ &= iub + \int \left( e^{iux} - 1 - iuxI(|x| \leq 1) \right) \Lambda(dx), \\ & b \geq \int_0^1 x \Lambda(dx) \text{ and Lévy measure } \Lambda \text{ concentrated on } (0, \infty). \end{aligned}$$

## CONCLUSION

Then along the same subsequence  $\{n'\}$ , as  $n' \rightarrow \infty$ ,

$$\left( \frac{\sum_{i=1}^{n'} X_i Y_i}{a_{n'}}, \frac{\sum_{i=1}^{n'} Y_i}{a_{n'}} \right) \xrightarrow{D} (W_1, W_2),$$

where

$$(W_1, W_2) \stackrel{D}{=} (a_1 + U, a_2 + V)$$

with  $(a_1, a_2) =$

$$\left( \left( b - \int_0^1 x \Lambda(dx) \right) \mathbb{E}X, b - \int_0^1 x \Lambda(dx) \right)$$

and  $\mathbb{E}e^{i(\theta_1 U + \theta_2 V)} =$

$$\begin{aligned} & \exp \left\{ \int_0^\infty \int_{-\infty}^\infty \left( e^{i(\theta_1 x + \theta_2 y)} - 1 \right) dF(x/y) \Lambda(dy) \right\} \\ & =: \exp \{ \phi(\theta_1, \theta_2) \}. \end{aligned}$$

# IMPLICATION

We see that whenever

$$\frac{1}{a_{n'}} \sum_{i=1}^{n'} Y_i \xrightarrow{D} W_2, \text{ as } n' \rightarrow \infty,$$

and  $P \{W_2 > 0\} = 1$ , then as  $n' \rightarrow \infty$

$$\mathbb{T}_{n'} = \sum_{i=1}^{n'} X_i Y_i / \sum_{i=1}^{n'} Y_i \xrightarrow{D} W_1 / W_2.$$

In particular this happens when  $Y$  is in the centered Feller class, since in this case necessarily  $W_2$  has a Lebesgue density.

## FELLER CLASS

$\xi$  is in the Feller class if for the sequence of partial sums  $\{S_n\}_{n \geq 1}$  of i.i.d.  $\xi$  there exist norming and centering constants  $B(n) > 0$ ,  $A(n)$  such that every subsequence of  $\{n_k\}$  of  $\{n\}$  contains a further subsequence  $n_{k'} \rightarrow \infty$  with

$$\frac{S_{n_{k'}} - A(n_{k'})}{B(n_{k'})} \xrightarrow{D} W,$$

where  $W$  is a finite nondegenerate rv depending on the subsequence  $n_{k'}$ , which by a result of Pruitt (1983) has a Lebesgue density.

We shall write this as “ $\xi \in FC$ ”.



## CENTERED FELLER CLASSES

If the centering function  $A(n)$  can be chosen to be identically equal to zero, we shall say:

$\xi$  is in the centered Feller class, written “ $\xi \in FC_0$ ”.

## THE CASE $Y \in FC_0$

Whenever  $Y \in FC_0$  one can show that  $a_1 = a_2 = 0$  and thus all of the subsequential distributional limits of  $\mathbb{T}_n$  are of the form  $U/V$  with  $P\{V > 0\} = 1$ .

Moreover, by using a result of Griffin (1986) it can be shown that these distributional limits have Lebesgue densities on  $\mathbb{R}$ .

## GOING THE OTHER WAY

Going the other way, Kevei and M (2013) show that if all of the subsequential laws of  $\mathbb{T}_n$  are continuous for any choice of  $X \in \mathcal{X}$ , then necessarily  $Y \in FC_0$ .

In particular, if  $Y \in FC$ , but  $Y \notin FC_0$  then there exists a subsequence  $\{n'\}$  such that

$$\sum_{i=1}^{n'} X_i Y_i / \sum_{i=1}^{n'} Y_i \xrightarrow{P} \mathbb{E}X, \text{ as } n' \rightarrow \infty.$$

# ASSOCIATED BIVARIATE LÉVY PROCESS

Under the assumptions of our basic result we can readily prove that for any  $t > 0$ , as  $n' \rightarrow \infty$ ,

$$\left( \frac{\sum_{1 \leq i \leq n't} X_i Y_i}{a_{n'}}, \frac{\sum_{1 \leq i \leq n't} Y_i}{a_{n'}} \right) \xrightarrow{D} (a_1 t + U_t, a_2 t + V_t),$$

where  $(U_t, V_t)$ ,  $t \geq 0$ , is the bivariate Lévy process with characteristic function

$$\mathbb{E} e^{i(\theta_1 U_t + \theta_2 V_t)} =: \exp \{t \phi(\theta_1, \theta_2)\}.$$

# ASYMPTOTIC DISTRIBUTION OF RATIO

Kevei and M (2013) have characterized when under regularity conditions that the ratio

$$T_t := U_t/V_t \xrightarrow{D} T.$$

converges in distribution to a nondegenerate random variable  $T$  as  $t \rightarrow \infty$  or  $t \searrow 0$ .

## KEVEI-M RESULT

They obtained the following analog for  $T_t$  of the MZ result:

Suppose  $X$  is non-degenerate and satisfies  $\mathbb{E}|X|^p < \infty$  for some  $p > 2$ , then the ratio as  $t \rightarrow \infty$  ( $t \searrow 0$ )

$$T_t := U_t/V_t \xrightarrow{D} T, \text{ nondegenerate,}$$

(in the case  $t \downarrow 0$  we assume  $\bar{\Lambda}(0+) = \infty$ .) if and only if for some  $0 \leq \beta < 1$ ,

$\bar{\Lambda}(x)$  is regularly varying at infinity (zero) with index  $-\beta$ .

# LEVY PROCESS FELLER CLASS

A Lévy process  $Y_t$  is said to be in the *Feller class* at infinity if there exists a norming function  $B(t)$  and a centering function  $A(t)$  such that for each sequence  $t_k \rightarrow \infty$  there exists a subsequence  $t'_k \rightarrow \infty$  such that

$$\left( Y_{t'_k} - A(t'_k) \right) / B(t'_k) \xrightarrow{D} W, \quad \text{as } k \rightarrow \infty,$$

where  $W$  is a nondegenerate random variable.

For the definition of *Feller class* at zero replace  $t_k \rightarrow \infty$  and  $t'_k \rightarrow \infty$ , by  $t_k \searrow 0$  and  $t'_k \searrow 0$ , respectively.

# LEVY PROCESS CENTERED FELLER CLASS

The Lévy process  $Y_t$  belongs to the *centered Feller class* at infinity if it is in the Feller class at infinity and the centering function  $A(t)$  can be chosen to be identically zero.

For the definitions of *centered Feller class* at zero replace  $t_k \rightarrow \infty$  and  $t'_k \rightarrow \infty$ , by  $t_k \searrow 0$  and  $t'_k \searrow 0$ , respectively.



## CONTINUOUS LIMITS

**Theorem** *All subsequential distributional limits of  $U_t/V_t$ , as  $t \searrow 0$ , (as  $t \rightarrow \infty$ ) are continuous for any cdf  $F$  in the class  $\mathcal{X}$ , if and only if  $V_t$  is in the centered Feller class at 0 ( $\infty$ ).*

## REPRESENTATION

Now let  $\{X_s\}_{s \geq 0}$  be a class of i.i.d.  $F$  random variables independent of the  $V_t$  process. It turns out that for each  $t \geq 0$  the bivariate process

$$(U_t, V_t) \stackrel{D}{=} \left( \sum_{0 \leq s \leq t} X_s \Delta V_s, \sum_{0 \leq s \leq t} \Delta V_s \right), \quad (6)$$

where  $\Delta V_s = V_s - V_{s-}$ .

Notice that in the representation (6) each jump of  $V_t$  is weighted by an independent  $X_t$  so that  $U_t$  can be viewed as a randomly weighted Lévy process.

## A PICTURE OF THIS PROCESS

Here is a graphic way to picture this bivariate process. Consider  $\Delta V_s$  as the intensity of a random shock to a system at time  $s > 0$  and  $X_s \Delta V_s$  as the cost of repairing the damage that it causes.

Then  $V_t$ ,  $U_t$  and  $U_t/V_t$  represent, respectively, up to time  $t$ , the total intensity of the shocks, the total cost of repair and the average cost of repair with respect to shock intensity.

## A MOTIVATING EXAMPLE

Let  $\Delta V_s$  represent a measure of the intensity of a tornado that comes down in a Midwestern American state at time  $s$  during tornado season and  $X_s$  the cost of the repair of the damage per intensity that it causes.

Note that  $X_s$  is a random variable that depends on where the tornado hits the ground, say a large city, a medium size town, a village, an open field, etc.

It is assumed that a tornado is equally likely to strike anywhere in the state.