

Financial Mathematics

Preliminaries to
the Triangular Central Limit Theorem
and a first proof of Black-Scholes

Fact (Second order Maclaurin approximations):
 f'' is continuous at 0 $\Rightarrow \exists \varepsilon(x) \rightarrow 0$ s.t.

$$f(x) = [f(0)] + [f'(0)]x + \left[\frac{f''(0)}{2} \right] x^2 + \boxed{[\varepsilon(x)]x^2}.$$

error term
 $o(x^2)$

Fact (First order Maclaurin approximations):
 f' is continuous at 0 $\Rightarrow \exists \varepsilon(x) \rightarrow 0$ s.t.
 $\underbrace{\hspace{10em}}_{\text{as } x \rightarrow 0}$

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 f'' is continuous at 0 $\Rightarrow \exists \varepsilon(x) \rightarrow 0$ s.t.

$$f(x) = [f(0)] + [f'(0)]x + \left[\frac{f''(0)}{2} \right] x^2 + [\varepsilon(x)]x^2.$$

Special case: $e^x = 1 + x + \frac{x^2}{2} + [\varepsilon(x)]x^2$

Corollary: $x_n \rightarrow 0 \Rightarrow \exists \varepsilon_n \rightarrow 0$ s.t. $e^{x_n} = 1 + x_n + \frac{x_n^2}{2} + \varepsilon_n x_n^2$

$x \rightarrow x_n, \varepsilon_n := \varepsilon(x_n) \xrightarrow{\text{as } n \rightarrow \infty} 0,$

$x_n \rightarrow 3x_n$

Corollary: $x_n \rightarrow 0 \Rightarrow \exists \varepsilon_n \rightarrow 0$ s.t. $e^{3x_n} = 1 + 3x_n + \frac{9x_n^2}{2} + 9\varepsilon_n x_n^2$

$9\varepsilon_n \rightarrow \delta_n$

i.e.: $\exists \delta_n \rightarrow 0$ s.t. $e^{3x_n} = 1 + 3x_n + \frac{9x_n^2}{2} + \delta_n x_n^2$

“ $9[o(x_n^2)] = o(x_n^2)$ ”

$o(x_n^2)$

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Def'n: $\boxed{h(x)} := e^{-x^2/2} / \sqrt{2\pi}$

Def'n: $Z_n \rightarrow Z$ in distribution means:

for any contin., bounded $\phi : \mathbb{R} \rightarrow \mathbb{R}$,

$$E[\phi(Z_n)] \rightarrow E[\phi(\boxed{Z})]. \quad \begin{array}{l} Z \text{ not} \\ \text{def'd, so...} \end{array}$$

Replace $E[\phi(Z)]$

$$\text{by } \int_{-\infty}^{\infty} [\phi(x)][h(x)] dx$$

“Change every Z to x
and then integrate against $h(x) dx$,
from $-\infty$ to ∞ .”

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Def'n: $\boxed{h(x)} := e^{-x^2/2} / \sqrt{2\pi}$

Def'n: $Z_n \rightarrow Z$ in distribution means:
equivalently: continuous, compactly supported
for any ~~contin., bounded~~ $\phi : \mathbb{R} \rightarrow \mathbb{R}$,

$$E[\phi(Z_n)] \rightarrow \int_{-\infty}^{\infty} [\phi(x)][h(x)] dx.$$

Fact: $Z_n \rightarrow Z$ in distribution
 $\alpha_n \rightarrow 0$ in \mathbb{R}
 $\Rightarrow Z_n + \alpha_n \rightarrow Z$ in distribution.
pf omitted

Fact: $Z_n \rightarrow Z$ in distribution
 $\alpha_n \rightarrow 1$ in \mathbb{R}
 $\Rightarrow \alpha_n Z_n \rightarrow Z$ in distribution.
pf omitted

Def'n: $\boxed{h(x)} := e^{-x^2/2} / \sqrt{2\pi}$

Def'n: $Z_n \rightarrow Z$ in distribution against contin, exp-bdd means:
continuous, exponentially bounded
for any ~~contin., bounded~~ $\phi : \mathbb{R} \rightarrow \mathbb{R}$,

$$E[\phi(Z_n)] \rightarrow \int_{-\infty}^{\infty} [\phi(x)][h(x)] dx.$$

Fact: $Z_n \rightarrow Z$ in distribution against contin, exp-bdd
 $\alpha_n \rightarrow 0$ in \mathbb{R}
 $\Rightarrow Z_n + \alpha_n \rightarrow Z$ in distribution! against contin, exp-bdd
pf omitted

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pf omitted

$D_{1/7}$:

$$\begin{array}{l} 0.5 \\ 0.5 \end{array} \left| \begin{array}{l} 1/7 \\ -1/7 \end{array} \right. \begin{array}{l} z^{1/7} \\ z^{-1/7} \end{array}$$

What about $D_{1/7}$?
Replace t by $t/7$.

$$i = \sqrt{-1}$$

Replace z by e^{-it}

$$(0.5)z^{1/7} + (0.5)z^{-1/7}$$

$$(0.5)e^{-it/7} + (0.5)e^{it/7}$$

$$\begin{array}{c} \parallel \\ \cos(t/7) \end{array}$$

$$\begin{aligned} e^{it/7} &= \cos(t/7) + i \sin(t/7) \\ e^{-it/7} &= \cos(t/7) - i \sin(t/7) \end{aligned}$$

$D_{1/7} :$

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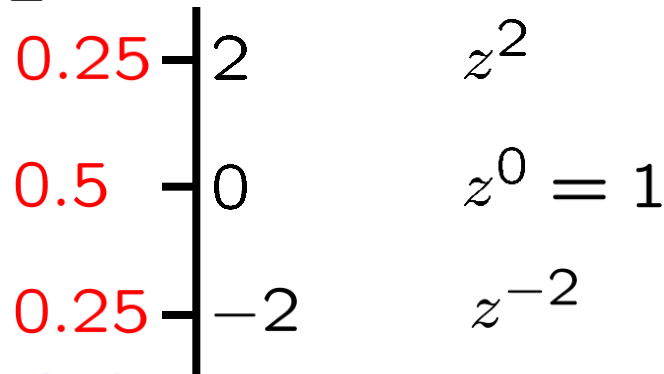
Generating function:

Fourier transform:

The Fourier transf. of the distr. of $X/7$
is equal to

(the Fourier transf. of the distr. of X)
 $t: \rightarrow t/7$

$$D_2 = H_2 - T_2 :$$



forget its origin keep the distribution

Generating function:

$$\begin{aligned} & (0.25)z^2 + 0.5 + (0.25)z^{-2} \\ & = ((0.5)z + (0.5)z^{-1})^2 \end{aligned}$$

the generating function
of the distribution
of D_1

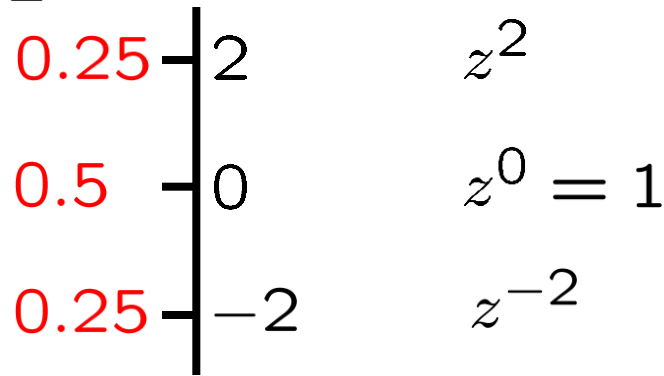
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Fourier transform:

$$(\cos t)^2 = \cos^2 t$$

$$D_2 = H_2 - T_2 :$$

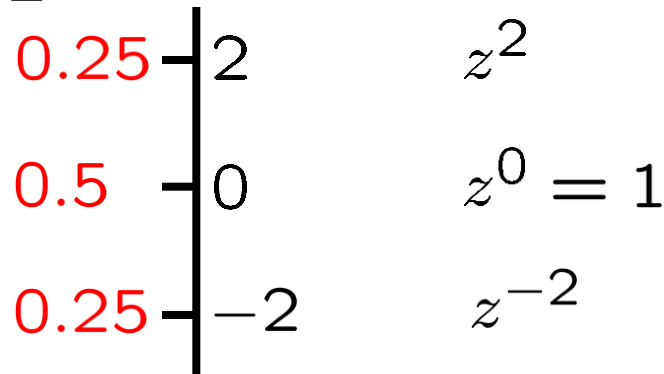


Fourier transform: $(\cos t)^2 = \cos^2 t$

Note: D_2 is a sum of two independent PCRVs.

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Fourier transform: $(\cos t)^2 = \cos^2 t$

Note: D_2 is a sum of two independent PCRVs. The Fourier transf. of the distr. of either one is $\cos t$.

X and Y independent \Rightarrow
 the Fourier transf. of the distr. of $X + Y$
 is the product of
 the Fourier transf. of the distr. of X
 and
 the Fourier transf. of the distr. of Y .

Computing the distribution of $X + Y$
from the distributions of X and Y
is generally **impossible**;
you need the JOINT distribution of X and Y .
If X and Y are independent,
then the joint distribution of X and Y
is the “product” of
the distributions of X and Y ,
and the distribution of $X + Y$
can be obtained from
the distributions of X and Y ,
by a complicated process called “convolution”.

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More on all that in a later lecture...

For now, just remember:

Fourier transf. simplifies convolution
to multiplication.

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Z : $\frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx$ $\Bigg|_x z^x$ Do this for all $x \in \mathbb{R}$

Generating function:

$$\int_{-\infty}^{\infty} z^x \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx = \text{Exercise}$$

Fourier transform:

$$\int_{-\infty}^{\infty} e^{-itx} \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx = e^{-t^2/2}$$

Key idea of Central Limit Theorem:

Let Z have distr. with Fourier transf. $e^{-t^2/2}$.
 Then Z is "close" to X . because: (F. transf. of distr. of X) $\approx e^{-t^2/2}$

$$Z: \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx \quad \Bigg| \quad x \quad z^x$$

Do this for
all $x \in \mathbb{R}$

If the Fourier transforms of the distributions of PCRVs X_1, X_2, \dots approach $e^{-t^2/2}$, then the distributions of X_1, X_2, \dots approach the distribution shown above, i.e., $X_n \rightarrow Z$ in distribution.

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Def'n: For any PCRV Y ,

$$\boxed{\mathcal{F}\delta[Y]} = \boxed{\mathcal{F}\delta_Y} := E[e^{-itY}] \stackrel{=} {=} \left[E[z^Y] \right]_{z \rightarrow e^{-it}}$$

is the **Fourier transform** \mathcal{F} of the δ distribution of Y .

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Key idea of Central Limit Theorem:

Let Z have distr. with Fourier transf. $e^{-t^2/2}$. Then Z is "close" to X . because: (F. transf. of distr. of X) $\approx e^{-t^2/2}$

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is the **Fourier transform** \mathcal{F} of the δ distribution of Y .

Fact: $\mathcal{F}\delta_{X/7} = [\mathcal{F}\delta_X]_{t:\rightarrow t/7}$

Fact: $\mathcal{F}\delta_{cX} = [\mathcal{F}\delta_X]_{t:\rightarrow ct}$

Fact: X_1, \dots, X_n indep. \Rightarrow
 $\mathcal{F}\delta_{X_1+\dots+X_n} = [\mathcal{F}\delta_{X_1}] \cdots [\mathcal{F}\delta_{X_n}]$

$$\mathcal{F}\delta_Z = e^{-t^2/2}$$

$$\mathcal{F}\delta_{Y_n} \xrightarrow{\text{Apply } \mathcal{F}^{-1}} \mathcal{F}\delta_Z$$

$$\delta_{Y_n} \xrightarrow{\text{Apply } \mathcal{F}^{-1}} \delta_Z$$

QED

Fact: $\mathcal{F}\delta_{Y_n} \xrightarrow{e^{-t^2/2}}$
 $\Rightarrow Y_n \rightarrow Z$ in distribution.

pf omitted, but here's the idea...

Def'n: Let \mathcal{A} be a set of PCRVs.

We say \mathcal{A} is **identically distributed** or **i.d.**

if $\forall A, B \in \mathcal{A}$,

A and B have the same distribution.

Def'n: If \mathcal{A} is i.d., then alternate: $\mathcal{F}\delta_{\mathcal{A}}$
 $E[\mathcal{A}]$, $\text{Var}[\mathcal{A}]$, $\text{SD}[\mathcal{A}]$, $\mathcal{F}\delta[\mathcal{A}]$
 are defined as alternate: $\mathcal{F}\delta_{\mathcal{A}}$
 $E[A]$, $\text{Var}[A]$, $\text{SD}[A]$, $\mathcal{F}\delta[A]$, resp.
 for any $A \in \mathcal{A}$.

Def'n: \forall set \mathcal{A} of PCRVs, $\forall c \in \mathbb{R}$,
 $c\mathcal{A} := \{cA \mid A \in \mathcal{A}\}$, $c + \mathcal{A} := \{c + A \mid A \in \mathcal{A}\}$

Fact: If \mathcal{A} is i.d., then $c\mathcal{A}$ is i.d., $c + \mathcal{A}$?
 and $E[c\mathcal{A}] = c(E[\mathcal{A}])$
 $\text{Var}[c\mathcal{A}] = c^2(\text{Var}[\mathcal{A}])$
 $\text{SD}[c\mathcal{A}] = |c|(\text{SD}[\mathcal{A}])$.

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Fact: If \mathcal{A} is i.d., then $c + \mathcal{A}$ is i.d.,
 and $E[c + \mathcal{A}] = c + (E[\mathcal{A}])$
 $\text{Var}[c + \mathcal{A}] = \text{Var}[\mathcal{A}]$
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Def'n: $\mathcal{S} := \{\text{standard PCRVs}\}$
 $= \{\text{PCRVs } X \mid E[X] = 0 \ \& \ \text{Var}[X] = 1\}$

Exercise: Show that \mathcal{S} is not i.d.

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Exercise: Show that \mathcal{S} is **not** i.d.

Hint: Find two standard binary PCRVs with different distributions.

Fact: If A is i.d., then

$$A \subseteq \mathcal{S} \iff E[A] = 0 \ \& \ \text{Var}[A] = 1.$$

Fact: If A is i.d., and if $A \cap \mathcal{S} \neq \emptyset$, then $A \subseteq \mathcal{S}$.

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Def'n: For any set \mathcal{A} of PCRVs,
let $\sum^n \mathcal{A}$ denote the set of all $A_1 + \dots + A_n$
such that $A_1, \dots, A_n \in \mathcal{A}$
and such that A_1, \dots, A_n are i.i.d.

Fact: If \mathcal{A} is i.d., and if $n \geq 1$ is an integer
then $\sum^n \mathcal{A}$ is i.d.,

Fact: If \mathcal{A} is i.d., then
 $\mathcal{A} \subseteq \mathcal{S} \iff E[\mathcal{A}] = 0$ and $\text{Var}[\mathcal{A}] = 1$.

Fact: If \mathcal{A} is i.d., and if $\mathcal{A} \cap \mathcal{S} \neq \emptyset$, then $\mathcal{A} \subseteq \mathcal{S}$.

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Fact: If \mathcal{A} is i.d., and if $n \geq 1$ is an integer
 then $\sum^n \mathcal{A}$ is i.d.,
 and $E[\sum^n \mathcal{A}] = n(E[\mathcal{A}])$ ← DIVIDE BY \sqrt{n}
 $\text{Var}[\sum^n \mathcal{A}] = n(\text{Var}[\mathcal{A}])$ ← DIVIDE BY n
 $\text{SD}[\sum^n \mathcal{A}] = \sqrt{n}(\text{SD}[\mathcal{A}])$
 $\mathcal{F}\delta[\sum^n \mathcal{A}] = (\mathcal{F}\delta[\mathcal{A}])^n$.

Fact: $\sum^n (\frac{c}{n} + \mathcal{A}) = c + (\sum^n \mathcal{A})$

Fact: $\sum^n (c\mathcal{A}) = c(\sum^n \mathcal{A})$
 $E[(\sum^n \mathcal{A})/\sqrt{n}] = \sqrt{n}(E[\mathcal{A}])$
 $\text{Var}[(\sum^n \mathcal{A})/\sqrt{n}] = \text{Var}[\mathcal{A}]$

Fact: $\mathcal{A} \subseteq \mathcal{S} \iff [\sum^n \mathcal{A}]/\sqrt{n} \subseteq \mathcal{S}$

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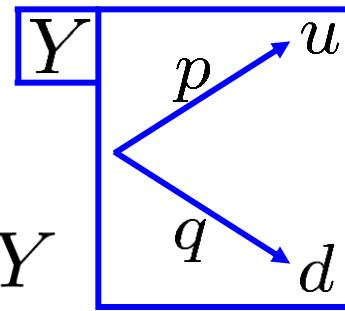
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renormalized

i.i.d. sum preserves and reflects standardness

Def'n: $\forall p, q \in [0, 1], \forall u, d \in \mathbb{R},$
 s.t. $p + q = 1$ s.t. $d < u$



let $\mathcal{B}_{q,d}^{p,u}$ be the set of binary PCRVSs Y
 such that $\Pr[Y = u] = p$ and $\Pr[Y = d] = q$.

Fact: $\mathcal{B}_{q,d}^{p,u}$ is i.i.d.

Fact: $c \neq 0 \Rightarrow c\mathcal{B}_{q,d}^{p,u} = \mathcal{B}_{q,cd}^{p,cu}$

Fact: $c + \mathcal{B}_{q,d}^{p,u} = \mathcal{B}_{q,d+c}^{p,u+c}$

Fact: $\sum^n (\frac{c}{n} + \mathcal{A}) = c + (\sum^n \mathcal{A})$

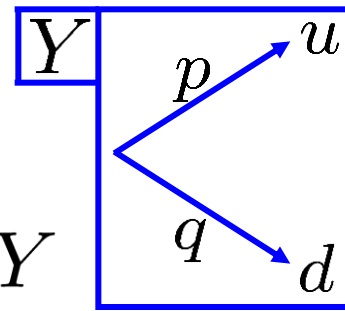
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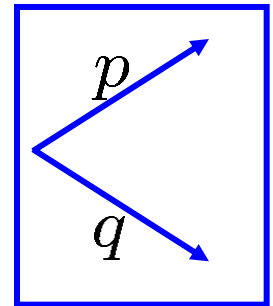
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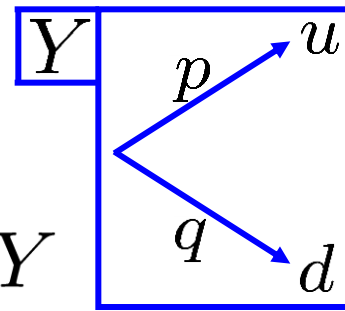
Def'n: $\forall p, q \in [0, 1],$
 s.t. $p + q = 1$ $\mathcal{B}_q^p := \bigcup_{d < u} \mathcal{B}_{q,d}^{p,u}$



$X_1, \dots, X_n \in \bigcup_{d < u} \mathcal{B}_{q,d}^{p,u}$ i.i.d. $\Rightarrow \exists (d < u)$ s.t. $X_1, \dots, X_n \in \mathcal{B}_{q,d}^{p,u}$

$$\sum^n \mathcal{B}_q^p = \sum^n \bigcup_{d < u} \mathcal{B}_{q,d}^{p,u} = \bigcup_{d < u} \sum^n \mathcal{B}_{q,d}^{p,u}$$

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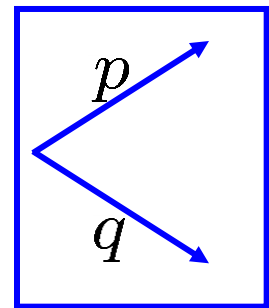
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Def'n: $\forall p, q \in [0, 1],$ s.t. $p + q = 1$, $\mathcal{B}_q^p := \bigcup_{d < u} \mathcal{B}_{q,d}^{p,u}$



The iid sum of general binaries is the union of the iid sums of specific binaries.

$$\sum^n \mathcal{B}_q^p = \sum^n \bigcup_{d < u} \mathcal{B}_{q,d}^{p,u} = \bigcup_{d < u} \sum^n \mathcal{B}_{q,d}^{p,u}$$

Lemma: If $X \in \sum^n \mathcal{B}_q^p$, $\alpha, \beta \in \mathbb{R}$, $\beta \neq 0$,
 then $\frac{X-\alpha}{\beta} \in \sum^n \mathcal{B}_q^p$.

Proof: Choose $u, d \in \mathbb{R}$

s.t. $X \in \sum^n \mathcal{B}_{q,d}^{p,u}$.

Then $X - \frac{\alpha}{n} \in \sum^n \left[\left(\mathcal{B}_{q,d}^{p,u} \right) - \frac{\alpha}{n} \right]$

Def'n: $\forall p, q \in [0, 1]$, $\mathcal{B}_q^p := \bigcup_{d < u} \mathcal{B}_{q,d}^{p,u}$
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$$\sum^n \mathcal{B}_q^p = \sum^n \bigcup_{d < u} \mathcal{B}_{q,d}^{p,u} = \bigcup_{d < u} \sum^n \mathcal{B}_{q,d}^{p,u}$$

Lemma: If $X \in \sum^n \mathcal{B}_q^p$, $\alpha, \beta \in \mathbb{R}$, $\beta \neq 0$,

then $\frac{X-\alpha}{\beta} \in \sum^n \mathcal{B}_q^p$.

Special case:

$$\alpha = \mathbb{E}[X],$$

$$\beta = \text{SD}[X].$$

Proof: Choose $u, d \in \mathbb{R}$

s.t. $X \in \sum^n \mathcal{B}_{q,d}^{p,u}$.

Then $X - \alpha \in \sum^n \left[\left(\mathcal{B}_{q,d}^{p,u} \right) - \frac{\alpha}{n} \right]$

$$= \sum^n \mathcal{B}_{q, d - (\alpha/n)}^{p, u - (\alpha/n)},$$

so $\frac{X-\alpha}{\beta} \in \sum^n \left[\frac{1}{\beta} \left(\mathcal{B}_{q, d - (\alpha/n)}^{p, u - (\alpha/n)} \right) \right]$

$$= \sum^n \mathcal{B}_{q, [d - (\alpha/n)]/\beta}^{p, [u - (\alpha/n)]/\beta} \subseteq \sum^n \mathcal{B}_q^p. \quad \text{QED}$$

Uptick/downtick VALUES change,
but uptick/downtick PROBABILITIES do not.

Lemma: If $X \in \Sigma^n \mathcal{B}_q^p$, $\alpha, \beta \in \mathbb{R}$, $\beta \neq 0$,

then $\frac{X-\alpha}{\beta} \in \Sigma^n \mathcal{B}_q^p$.

Special case:

$$\alpha = E[X],$$

$$\beta = SD[X].$$

Def'n: $X_\circ = \frac{X - (E[X])}{SD[X]}$

= the **standardization** of X

Lemma: If $X \in \Sigma^n \mathcal{B}_q^p$,

then $X_\circ \in \Sigma^n \mathcal{B}_q^p$.

Def'n: For any set \mathcal{A} of PCRVs,
 let $\sum^n \mathcal{A}$ denote the set of all $A_1 + \dots + A_n$
 such that $A_1, \dots, A_n \in \mathcal{A}$
 and such that A_1, \dots, A_n are i.i.d.

Def'n: For any set \mathcal{A} of PCRVs,
 let $\prod^n \mathcal{A}$ denote the set of all $A_1 \cdots A_n$
 such that $A_1, \dots, A_n \in \mathcal{A}$
 and such that A_1, \dots, A_n are i.i.d.

Def'n: $\exp(\mathcal{A}) := e^{\mathcal{A}} := \{e^A \mid A \in \mathcal{A}\}$.

Fact: $e^{\sum^n \mathcal{A}} = \prod^n e^{\mathcal{A}}$,
 i.e., $\exp(\sum^n \mathcal{A}) = \prod^n [\exp(\mathcal{A})]$

Fact: \mathcal{A} is i.d. $\Rightarrow \prod^n \mathcal{A}$ is i.d.
 and $E[\prod^n \mathcal{A}] = (E[\mathcal{A}])^n$.

