

# Large Deviations on Sets of Words

## ALEA'10

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# Motivation

Find exceptional words, assess the significance

- Problem statement

Compute  $P(X_n \geq k)$ , where  $X_n$  is the r.v. that counts occurrences of a set of words in a random text of size  $n$

- Method Use generating functions and combinatorial properties of words to compute LD results.

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- **Method** Use generating functions and combinatorial properties of words to **compute** LD results.

# Central Limit Theorem

## Theorem

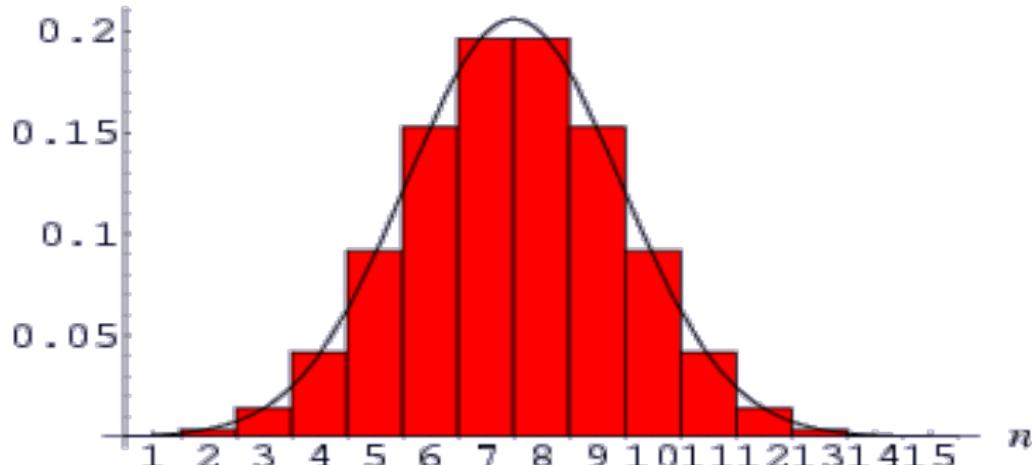
Let  $X_1, \dots, X_n$  be i.i.d. random variables with mean  $\mu$  and variance  $\sigma^2$ , with  $0 < \sigma^2 < +\infty$ . Then

$$\frac{\sum_{i=1}^n X_i - n\mu}{\sqrt{n}\sigma} \rightarrow \mathcal{N}(0, 1)$$

when  $n \rightarrow \infty$ .

## Illustration

$$P_{0.5}(n | 15)$$



Zscore = comparison with normal law

$$Z(H) = \frac{O(H) - E(H)}{\sqrt{V(H)}}$$

$$Z(H) \rightarrow \mathcal{N}(0, 1)$$

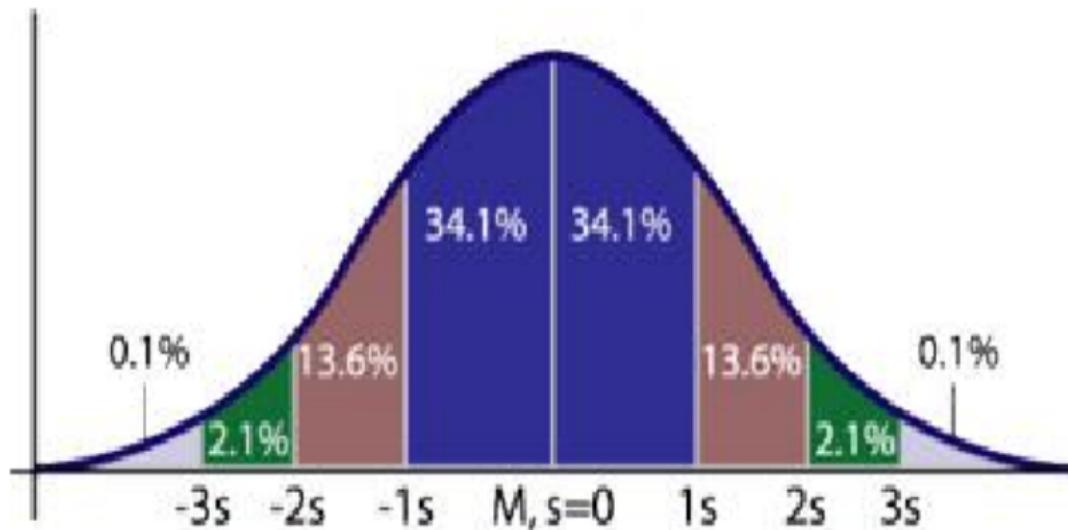
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## Exceptional events



# (Discrete) Probability generating functions

Def.

$$\begin{aligned}\phi_X(t) &= \sum_{k=0}^{\infty} e^{tk} P(X = k) , \\ \psi_X(u) &= \sum_{k=0}^{\infty} u^k P(X = k) = \phi(t = \log u) .\end{aligned}$$

Levy's Th.

Let  $(X_n)$  be a sequence of r.v. and  $X$  be a r.v. If

$$\phi_{X_n}(t) \rightarrow \phi_X(t) ,$$

when  $t$  is in a neighbourhood of 0, then  $X_n \rightarrow X$  (cv. in law).

# Large deviations: basics

## Law of Large numbers

$$S_n = \sum_i X_i, \quad X_i \text{ i.i.d.}$$

$$\forall \epsilon > 0 : \lim \text{Prob}\left(\left|\frac{S_n}{n} - \frac{E(S_n)}{n}\right| > \epsilon\right) \rightarrow 0 .$$

## Large deviation definitions

$$\frac{1}{\phi(n)} \log \text{Prob}(X_n \geq a > E(X_n)) \rightarrow I(a)$$

$\phi(n)$  =  $\sqrt{n}, n, \dots$  : *speed*

$I(a)$  : *rate*

Words: Combinatorial properties → explicit expression.

# Large deviations: basics

A direct computation: Bernoulli

$$P(X_n = na) = \binom{n}{na} p^{na} (1-p)^{n(1-a)}$$

Stirling formula

$$n! \sim e^{-n} n^{n+1/2} \sqrt{2\pi}$$

$\rightarrow \dots$

$$P(X_n = na) \sim \dots e^{-n(a \log \frac{a}{p} + (1-a) \log(\frac{1-a}{1-p}))}$$

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# LD: generating functions scheme

## Generating functions

$$\begin{aligned} P(X_n = na) &= [u^{na}] \psi_{X_n}(u) \\ (\text{Cauchy's formula}) &= \frac{1}{2i\pi} \int \frac{\psi_{X_n}(u)}{u^{na+1}} du \end{aligned}$$

Integrand  $e^{\log \phi_{X_n}(t) - nat} = e^{n[\frac{1}{n} \log \phi_{X_n}(t) - at]}$

$$h_a(t) = \frac{1}{n} \log \phi_{X_n}(t) - at.$$

Hint for the scheme

$$h_a(t) = h_a(t_a) + h'_a(t_a)(t - t_a) + O((t - t_a)^2)$$

$$\frac{1}{n} \log P(X_n = na) \rightarrow I(a) = h_a(t_a)$$

with:  $h'_a(t_a) = 0, n(t - t_a)^2 \geq 0.$

# Large deviations: simple examples

Bernoulli

$$\phi_{X_n}(t) = (1 + p(e^t - 1))^n ,$$

$$h_a(t) = \log[1 + p(e^t - 1)] - at$$

$$h'_a(t) = \frac{pe^t}{1 + p(e^t - 1)} - a$$

$$t_a = \log\left[\frac{a}{p} \cdot \frac{1-p}{1-a}\right]$$

$$h_a(t_a) = a \log \frac{a}{p} + (1-a) \log\left(\frac{1-a}{1-p}\right)$$

# Poisson

$$\begin{aligned}\phi_{X_n}(t) &= e^{\lambda(e^t - 1)} , \\ h_a(t) &= p(e^t - 1) - at \\ h'_a(t) &= pe^t - a\end{aligned}$$

$$\begin{aligned}t_a &= \log \frac{a}{p} \\ h_a(t_a) &= a - p - a \log\left(\frac{a}{p}\right)\end{aligned}$$

# Normal distribution

Mean  $np$ , variance  $n\sigma^2$ .

$$\phi_{X_n}(t) = e^{pt - \frac{\sigma^2 t^2}{2}},$$

$$h_a(t) = -\sigma^2 \frac{t^2}{2} + (p-a)t$$

$$h'_a(t) = -\sigma^2 t + (p-a)$$

$$t_a = \frac{p-a}{2\sigma^2}$$

$$h_a(t_a) = \frac{(p-a)^2}{2\sigma^2})$$

**Remark:**  $B, P \rightarrow \mathcal{N}(0, 1)$

large deviations are different...

# Large deviations: a single word

$$\begin{aligned} P(X_n \geq na) &= P(X \geq k) \\ L_k(z) &= \sum_{n \geq 0} P(X_n \geq k) z^n \\ &= R(z) \cdot M^{\textcolor{red}{na}-1}(z) \cdot \frac{1}{1-z} . \end{aligned}$$

- $R$ : s.g. 1st occurrence;
- $M$ : s.g. other occurrences (possibly overlapping);
- $\frac{1}{1-z}$ : s.g. all words.

## Large deviations: a single word (2)

$$\begin{aligned} P(X_n \geq na) &= \cdots \int \frac{L_k(z)}{z^{na+1}} dz \\ &\rightarrow \log R(z) \cdot M^{na-1}(z) \cdot \frac{1}{1-z} - (na+1) \log z \\ h_a(z) &= \log M(z) - a \log z . \end{aligned}$$

where

$$M(z) = 1 - \frac{A_H^{-1}(z)}{1 + \frac{P(H)z^{|H|}A_H^{-1}(z)}{1-z}} .$$

## Theorem

Let  $a > P(H)$ . Then:

$$\text{Prob}(N(H) \geq na) \sim \frac{1}{\sigma_a \sqrt{n}} e^{-nI(a) + \delta_a}$$

where

$$I(a) = a \ln \left( 1 - \frac{1 - z_a}{(1 - z_a)(1 - A(z_a)) - z^m P(H)} \right) + \ln z_a$$

and  $z_a$  is a solution of

$$\begin{aligned} 0 &= (1 - z)^2 [A(z)^2 - A(z) - azA'(z)] \\ &+ z^m P(H) \\ &\times [(1 - z)(2A(z) - am - 1) + z^m P(H) - az] . \end{aligned}$$

Complexity: Solve a polynomial equation  
vs exponential algorithm [Nuel01].

# Several words

$$M(z) \rightarrow \mathbb{M}(z)$$

Matrix Dimension  $|\mathcal{H}|$

$\mathbb{M}$  diagonal:  $\lambda_1, \dots, \lambda_{|\mathcal{H}|}$ .

$$\begin{aligned} L_k(z) &= (\alpha_1, \dots) \mathbb{M}^{k-1} \begin{pmatrix} \beta_1 \\ \vdots \\ \beta_k \end{pmatrix} \\ &= \alpha_1 \lambda_1(z)^{k-1} \beta_1 + \dots . \end{aligned}$$

$$h_a(z) = a \log \lambda_1(z) - \log z$$

# Large deviations: two words

Counting on two strands

- $\mathbb{M}(z)$  is a  $2 \times 2$  matrix;
- $d(z) = \text{determinant}(\mathbb{M}(z))$ ;
- $t(z) = \text{Trace}(\mathbb{M}(z))$ .

Dependency to overlaps

## Definition

Given a real  $a$ , the equation:

$$(azt'(z) - t(z))^2 d(z) + t(z)(azd'(z) - 2d(z))(t(z) - azt'(z)) + (azd'(z) - 2d(z))^2 = 0$$

is called the *fundamental equation*.

## Theorem

The rate function is :

$$I(a) = -a \log \lambda(z_a) + \log z_a \quad (1)$$

where

$$\lambda(z) = \frac{az(-\psi'(z)\theta(z) - (1 - \psi(z))\theta(z)) - 2(1 - \psi(z))\theta(z) - 2\theta(z)^2}{az(-\psi'(z)\theta(z) + \psi(z)\theta'(z)) - 2\theta(z)^2 + \psi(z)\theta(z)} \quad (2)$$

and  $z_a$  is the fundamental root of the fundamental equation.

## Hint for the proof

$$\lambda^2(z) - t(z)\lambda + d(z) = 0 \quad .$$

Rational expression for  $\lambda$ :

$$\lambda(z) = \frac{azd'(z) - 2d(z)}{azt'(z) - t(z)} \quad .$$

$$\lambda_1(z) = \frac{t(z) + \sqrt{\Delta(z)}}{2}$$

$$\lambda_2(z) = \frac{t(z) - \sqrt{\Delta(z)}}{2}$$

$$\lambda'_1(z) = \frac{1}{2}(t'(z) + \frac{1}{2} \frac{\Delta'(z)}{\sqrt{\Delta(z)}}) \quad .$$

Equation  $h'_a(z) = 0$  rewrites :  $az\lambda'_1(z) - \lambda_1(z) = 0$

$$\Rightarrow (azt'(z) - t(z))\sqrt{\Delta(z)} = \frac{2\Delta(z) - az\Delta'(z)}{2}$$

## Hint for the proof (2)

$$\lambda^2(z) - t(z)\lambda + d(z) = 0 \ .$$

$$(azt'(z) - t(z))\sqrt{\Delta(z)} = \frac{2\Delta(z) - az\Delta'(z)}{2}$$

$$\Delta(z) = t^2(z) - 4d(z) \rightarrow \Delta'(z) = \dots$$

$$\Delta(z) - \frac{az\Delta'(z)}{2} = (t(z) - azt'(z))t(z) + 2azd'(z) - 4d(z) \ .$$

Squaring  $\sqrt{\Delta}$ , substituting in expression for  $\lambda$ :

$$\lambda(z) = \frac{azd'(z) - 2d(z)}{azt'(z) - t(z)} \ .$$

# Comparison with normal approximation

Zscore theory

LD rates

$$I_N(a) = \frac{(a - p)^2}{2p}$$
$$I(a) \sim a \log \frac{a}{p}$$

Therefore

$$I(a) < I_N(a) : \text{underestimation}$$

$$I(a) \sim I_N(a) \Leftrightarrow a - p \ll p : \text{central domain}$$

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# Convergence paradox

Ratio and relative errors decrease exponentially

$$n \uparrow, e^{-nl(a)} \downarrow$$

## Total variation distance

- evaluates the area between two different distributions
- returns the largest probability.
- The smaller is the maximum distribution
- The bigger is the relative error
- The worse is the approximation

The normal approximation is never correct

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# Spouge's results

*Homo sapiens* promoter sequences: 30 most significant words sorted by p-value.

Word	ref.	occ.	p-val (rate)	Z- rank	Z-score	Normal-rate
TTTTTTTT	1	22589	0.000638191892	2	+264.507244	2.460797e-03
AAAAAAA	2	20828	0.0006320334298	1	+271.099279	2.584981e-03
GATTACAG	3	3149	0.0004108880766	3	+213.754122	1.607051e-03
GGATTACA	3b	3039	0.0004008417233	4	+212.430286	1.587207e-03
GGGATTAC	3c	2837	0.0003594976322	5	+196.306517	1.355408e-03
TGGGATTA	3d	3098	0.0003556183659	6	+188.531536	1.250169e-03
ATTACAGG	3e	3153	0.0003397024728	8	+174.786684	1.074527e-03
TGTAATCC	4	2560	0.0003092471632	7	+177.115531	1.103352e-03
CTGTAATC	4b	2579	0.0003038123568	9	+173.212685	1.055261e-03
TAATCCCA	4c	2610	0.000291167097	12	+164.545919	9.523023e-04
GTAATCCC	4e	2381	0.0002799502047	11	+165.967080	9.688232e-04
GCTGGGAT	5	3041	0.0002776774186	15	+146.473114	7.545994e-04
GTGTGTGT	6	3747	0.0002771092421	10	+171.091543	1.029574e-03
CAGGCTGG	7	4086	0.0002665394041	32	+127.732346	5.738553e-04
CCAGGCTG	7b	3931	0.0002512786464	33	+123.609400	5.374074e-04
CCTGTAAT	4b	2629	0.0002502132673	20	+141.622084	7.054441e-04
TGTGTGTG	6b	3866	0.0002437044783	16	+146.040997	7.501536e-04
CTGGGATT	8	3119	0.0002410460071	31	+128.161604	5.777188e-04
ACACACAC	9	3164	0.0002408138032	14	+160.598938	9.071643e-04

*B*Homo sapiens promoter sequences: next significant words sorted by p-value.

Word	ref.	occ.	p-val (rate)	Z- rank	Z-score	Normal-rate
ATCCCAGC	10	2623	0.0002316849672	27	+131.961279	6.124825e-04
CCAGCTG	11	3743	0.0002302858573	39	+116.774661	4.796206e-04
CCCAGCTA	10b	2428	0.0002296464312	26	+135.202979	6.429440e-04
AGTAGCTG	13	2048	0.000227794136	18	+145.360749	7.431816e-04
TTAGTACA	14	1672	0.0002273984643	13	+163.174975	9.364999e-04
TAGCTGGG	13b	2498	0.0002260538932	29	+131.682643	6.098987e-04
CAGCCTGG	11b	3741	0.0002258388149	42	+115.030518	4.654004e-04
CACACACA	9b	3242	0.0002097672137	23	+135.956731	6.501328e-04
CAGCTACT	11'd	1862	0.0002022195337	25	+135.290088	6.437727e-04
GTAGCTGG	13b	1968	0.0002021605972	28	+131.709404	6.101466e-04
TCAGCCTC	11c	2951	0.0002021588294	44	+113.055714	4.495579e-04

**Table:** The last column displays the rate function for normal approximation.

$n = 14215737$ .