

# Efficient Computation of Ratios of Stirling Numbers

Sara Kropf

Institute of Statistical Science, Academia Sinica

joint work with Hsien-Kuei Hwang

A3CS, Paris, September, 2, 2016



## Stirling Numbers (of the 2<sup>nd</sup> Kind)

$\left\{ \begin{matrix} n \\ k \end{matrix} \right\}$  = number of partitions of  $\{1, 2, \dots, n\}$  with  $k$  subsets  
= number of ways to nest  $n$  matryoshkas so you can still see  $k$



## Stirling Numbers (of the 2<sup>nd</sup> Kind)

$\left\{ \begin{matrix} n \\ k \end{matrix} \right\}$  = number of partitions of  $\{1, 2, \dots, n\}$  with  $k$  subsets  
= number of ways to nest  $n$  matryoshkas so you can still see  $k$



$\{1, 2, 3\}$

$\{1\} \cup \{2, 3\}, \quad \{2\} \cup \{1, 3\}, \quad \{3\} \cup \{1, 2\}$

$\{1\} \cup \{2\} \cup \{3\}$

# Stirling Numbers (of the 2<sup>nd</sup> Kind)

$\left\{ \begin{matrix} n \\ k \end{matrix} \right\}$  = number of partitions of  $\{1, 2, \dots, n\}$  with  $k$  subsets  
= number of ways to nest  $n$  matryoshkas so you can still see  $k$



$n$	$k = 1$	2	3	4	5	6
1	1					
2	1	1				
3	1	3	1			
4	1	7	6	1		
5	1	15	25	10	1	
6	1	31	90	65	15	1
$\vdots$						

$\{1, 2, 3\}$

$\{1\} \cup \{2, 3\}, \quad \{2\} \cup \{1, 3\}, \quad \{3\} \cup \{1, 2\}$

$\{1\} \cup \{2\} \cup \{3\}$

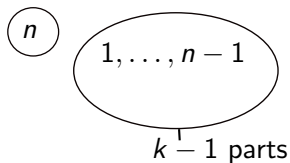


## Uniform Sampling of Partitions

$$\left\{ \begin{matrix} n \\ k \end{matrix} \right\} = \left\{ \begin{matrix} n-1 \\ k-1 \end{matrix} \right\} + k \left\{ \begin{matrix} n-1 \\ k \end{matrix} \right\}$$

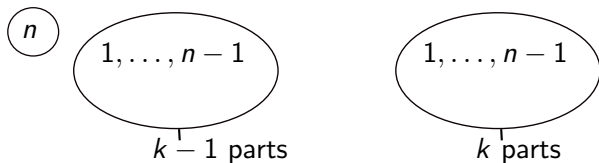
# Uniform Sampling of Partitions

$$\begin{Bmatrix} n \\ k \end{Bmatrix} = \begin{Bmatrix} n-1 \\ k-1 \end{Bmatrix} + k \begin{Bmatrix} n-1 \\ k \end{Bmatrix}$$



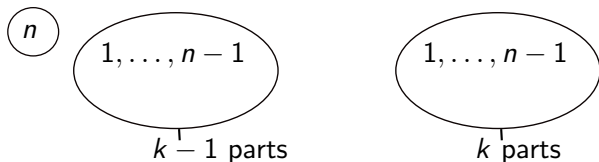
# Uniform Sampling of Partitions

$$\begin{Bmatrix} n \\ k \end{Bmatrix} = \begin{Bmatrix} n-1 \\ k-1 \end{Bmatrix} + k \begin{Bmatrix} n-1 \\ k \end{Bmatrix}$$



# Uniform Sampling of Partitions

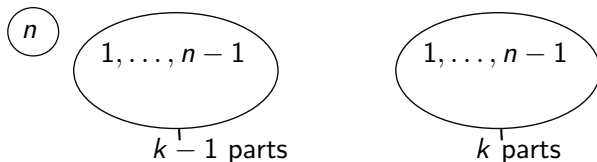
$$1 = \frac{\binom{n-1}{k-1}}{\binom{n}{k}} + \frac{k \binom{n-1}{k}}{\binom{n}{k}}$$





# Uniform Sampling of Partitions

$$1 = \frac{\binom{n-1}{k-1}}{\binom{n}{k}} + \frac{k \binom{n-1}{k}}{\binom{n}{k}}$$



Procedure for a uniform partition of  $\{1, \dots, n\}$  with  $k$  parts:

- Bernoulli RV  $X_{n,k}$  with  $\mathbb{P}(X_{n,k} = 1) = \frac{k \binom{n-1}{k}}{\binom{n}{k}}$
- If  $X_{n,k} = 1$ , then sample a partition of  $\{1, \dots, n-1\}$  with  $k$  parts and add the  $n$ -th element to one
- Otherwise sample a partition of  $\{1, \dots, n-1\}$  with  $k-1$  parts



# Number of Classes in a Population

- Population of unknown size
- Partitioned into  $\theta$  distinct classes (equally likely)



# Number of Classes in a Population

- Population of unknown size
- Partitioned into  $\theta$  distinct classes (equally likely)
- Draw a random sample of size  $n$  (multinomially distributed)
- We observe  $k$  different classes



# Number of Classes in a Population

- Population of unknown size
- Partitioned into  $\theta$  distinct classes (equally likely)
- Draw a random sample of size  $n$  (multinomially distributed)
- We observe  $k$  different classes
- Minimum variance unbiased estimator (MVUE) for  $\theta$  is (Charalambides 1968)

$$\frac{\binom{n+1}{k}}{\binom{n}{k}}$$

if  $n \geq \theta$



# Ratios

- Monotonicity, concavity and convexity
- Convergence of series (ratio test)
- Statistical tests (likelihood ratio test, variance ratio test)
- Conditional probability, correlation coefficient
- Optimality of heuristics
- ...



# Ratios

- Monotonicity, concavity and convexity
  - Convergence of series (ratio test)
  - Statistical tests (likelihood ratio test, variance ratio test)
  - Conditional probability, correlation coefficient
  - Optimality of heuristics
  - ...
- 
- Fast, efficient computation
  - Precise results
  - For two-parameter ratios: uniform result



## First Example

Using Stirling's formula:

$$\frac{\Gamma(n+x)}{\Gamma(n)} \sim \frac{\sqrt{\frac{2\pi}{n+x}} \left(\frac{n+x}{e}\right)^{n+x}}{\sqrt{\frac{2\pi}{n}} \left(\frac{n}{e}\right)^n} \sim n^x$$

## First Example

Using Stirling's formula:

$$\frac{\Gamma(n+x)}{\Gamma(n)} \sim \frac{\sqrt{\frac{2\pi}{n+x}} \left(\frac{n+x}{e}\right)^{n+x}}{\sqrt{\frac{2\pi}{n}} \left(\frac{n}{e}\right)^n} \sim n^x$$

Is there a cancellation-free approach?





## First Example

Using Stirling's formula:

$$\frac{\Gamma(n+x)}{\Gamma(n)} \sim \frac{\sqrt{\frac{2\pi}{n+x}} \left(\frac{n+x}{e}\right)^{n+x}}{\sqrt{\frac{2\pi}{n}} \left(\frac{n}{e}\right)^n} \sim n^x$$

Is there a cancellation-free approach?

$$\Gamma(n+x) = \int_0^\infty e^{-t} t^{n-1} g(t) dt \quad \text{with } g(t) = t^x$$



## First Example

Using Stirling's formula:

$$\frac{\Gamma(n+x)}{\Gamma(n)} \sim \frac{\sqrt{\frac{2\pi}{n+x}} \left(\frac{n+x}{e}\right)^{n+x}}{\sqrt{\frac{2\pi}{n}} \left(\frac{n}{e}\right)^n} \sim n^x$$

Is there a cancellation-free approach?

$$\Gamma(n+x) = \int_0^\infty e^{-t} t^{n-1} g(t) dt \quad \text{with } g(t) = t^x$$

Expand  $g(t)$  at  $t = n$ :

$$\Gamma(n+x) = \int_0^\infty e^{-t} t^{n-1} (n^x + \dots) dt = n^x \Gamma(n) + \dots$$



# Stirling Number

Recursion:

$$\left\{ \begin{matrix} n \\ k \end{matrix} \right\} = \left\{ \begin{matrix} n-1 \\ k-1 \end{matrix} \right\} + k \left\{ \begin{matrix} n-1 \\ k \end{matrix} \right\}$$

# Stirling Number

Recursion:

$$\left\{ \begin{matrix} n \\ k \end{matrix} \right\} = \left\{ \begin{matrix} n-1 \\ k-1 \end{matrix} \right\} + k \left\{ \begin{matrix} n-1 \\ k \end{matrix} \right\}$$

Extraction of a coefficient:

$$\left\{ \begin{matrix} n \\ k \end{matrix} \right\} = \frac{n!}{k!} [z^n] (e^z - 1)^k$$



# Stirling Number

Recursion:

$$\left\{ \begin{matrix} n \\ k \end{matrix} \right\} = \left\{ \begin{matrix} n-1 \\ k-1 \end{matrix} \right\} + k \left\{ \begin{matrix} n-1 \\ k \end{matrix} \right\}$$

Extraction of a coefficient:

$$\left\{ \begin{matrix} n \\ k \end{matrix} \right\} = \frac{n!}{k!} [z^n] (e^z - 1)^k$$

- Both are quite difficult to compute.
- There is no simple closed form expression.



## Growth of Stirling Numbers

For  $\frac{k}{n}$  in a closed subinterval of  $(0, 1)$ :

$$\left\{ \begin{matrix} n \\ k \end{matrix} \right\} \sim \frac{n!}{k!} \rho^{-n} (e^\rho - 1)^k \frac{1}{\sqrt{2\pi k \sigma^2}}$$

where  $\rho$  is the saddle point:

$$\frac{1 - e^{-\rho}}{\rho} = \frac{k}{n}$$

Similarly for  $\frac{k}{n}$  going to 0 or 1



## Growth of Stirling Numbers

For  $\frac{k}{n}$  in a closed subinterval of  $(0, 1)$ :

$$\left\{ \begin{matrix} n \\ k \end{matrix} \right\} \sim \frac{n!}{k!} \rho^{-n} (e^\rho - 1)^k \frac{1}{\sqrt{2\pi k \sigma^2}}$$

where  $\rho$  is the saddle point:

$$\frac{1 - e^{-\rho}}{\rho} = \frac{k}{n}$$

Similarly for  $\frac{k}{n}$  going to 0 or 1

We expect (see Harris 1968)

$$\frac{\left\{ \begin{matrix} n-1 \\ k \end{matrix} \right\}}{\left\{ \begin{matrix} n \\ k \end{matrix} \right\}} \sim \frac{\frac{(n-1)!}{k!} \rho^{-n+1} (e^\rho - 1)^k \frac{1}{\sqrt{2\pi k \sigma^2}}}{\frac{n!}{k!} \rho^{-n} (e^\rho - 1)^k \frac{1}{\sqrt{2\pi k \sigma^2}}} \sim \frac{\rho}{n}$$



## Growth of Stirling Numbers

For  $\frac{k}{n}$  in a closed subinterval of  $(0, 1)$ :

$$\left\{ \begin{matrix} n \\ k \end{matrix} \right\} \sim \frac{n!}{k!} \rho^{-n} (e^\rho - 1)^k \frac{1}{\sqrt{2\pi k \sigma^2}}$$

where  $\rho$  is the saddle point:

$$\frac{1 - e^{-\rho}}{\rho} = \frac{k}{n}$$

Similarly for  $\frac{k}{n}$  going to 0 or 1

We expect (see Harris 1968)

$$\frac{\left\{ \begin{matrix} n-1 \\ k \end{matrix} \right\}}{\left\{ \begin{matrix} n \\ k \end{matrix} \right\}} \sim \frac{\frac{(n-1)!}{k!} \rho^{-n+1} (e^\rho - 1)^k \frac{1}{\sqrt{2\pi k \sigma^2}}}{\frac{n!}{k!} \rho^{-n} (e^\rho - 1)^k \frac{1}{\sqrt{2\pi k \sigma^2}}} \sim \frac{\rho}{n}$$

Cancellations occur:

$$\frac{\text{Large}}{\text{Large}} \sim \text{Small}$$





# Stirling Numbers

- Growth of  $\left\{ \begin{matrix} n \\ k \end{matrix} \right\}$ :
  - Laplace 1814:  $k \asymp n$
  - Moser–Wyman 1958:  $n - k = o(\sqrt{n})$  and  $n - k \rightarrow \infty$
  - Temme 1993: uniform expansion for  $1 \leq k \leq n$
  - ...



# Stirling Numbers

- Growth of  $\left\{ \begin{matrix} n \\ k \end{matrix} \right\}$ :
  - Laplace 1814:  $k \asymp n$
  - Moser–Wyman 1958:  $n - k = o(\sqrt{n})$  and  $n - k \rightarrow \infty$
  - Temme 1993: uniform expansion for  $1 \leq k \leq n$
  - ...
- Ratios  $\frac{\left\{ \begin{matrix} n-1 \\ k \end{matrix} \right\}}{\left\{ \begin{matrix} n \\ k \end{matrix} \right\}}$ :
  - Ahuja 1972, Berg 1975: via recursions
  - Harris 1968, Hennecart 1994, Holst 1981: asymptotics of the first main term



## Direct Approach

$$\phi(z) = e^z - 1, \alpha = \frac{k}{n}$$



## Direct Approach

$$\phi(z) = e^z - 1, \alpha = \frac{k}{n}$$

Cauchy's integration formula

$$\begin{aligned} \left\{ \begin{matrix} n-1 \\ k \end{matrix} \right\} &= \frac{(n-1)!}{k!} [z^{n-1}] \phi(z)^k \\ &= \frac{(n-1)!}{k!} \frac{1}{2\pi i} \oint_{|z|=\rho} \phi(z)^k z^{-n-1} \cdot z \, dz \end{aligned}$$



## Direct Approach

$$\phi(z) = e^z - 1, \alpha = \frac{k}{n}$$

Cauchy's integration formula

$$\begin{aligned} \left\{ \begin{matrix} n-1 \\ k \end{matrix} \right\} &= \frac{(n-1)!}{k!} [z^{n-1}] \phi(z)^k \\ &= \frac{(n-1)!}{k!} \frac{1}{2\pi i} \oint_{|z|=\rho} \phi(z)^k z^{-n-1} \cdot z \, dz \end{aligned}$$

Taylor expansion of  $z = \rho + (z - \rho)$ :

$$\begin{aligned} \left\{ \begin{matrix} n-1 \\ k \end{matrix} \right\} &= \frac{(n-1)!}{k!} \frac{1}{2\pi i} \oint_{|z|=\rho} \phi(z)^k z^{-n-1} \, dz \rho \\ &\quad + \frac{(n-1)!}{k!} \frac{1}{2\pi i} \oint_{|z|=\rho} \phi(z)^k z^{-n-1} (z - \rho) \, dz \end{aligned}$$



## Direct Approach

$$\phi(z) = e^z - 1, \alpha = \frac{k}{n}$$

Cauchy's integration formula

$$\begin{aligned} \left\{ \begin{matrix} n-1 \\ k \end{matrix} \right\} &= \frac{(n-1)!}{k!} [z^{n-1}] \phi(z)^k \\ &= \frac{(n-1)!}{k!} \frac{1}{2\pi i} \oint_{|z|=\rho} \phi(z)^k z^{-n-1} \cdot z \, dz \end{aligned}$$

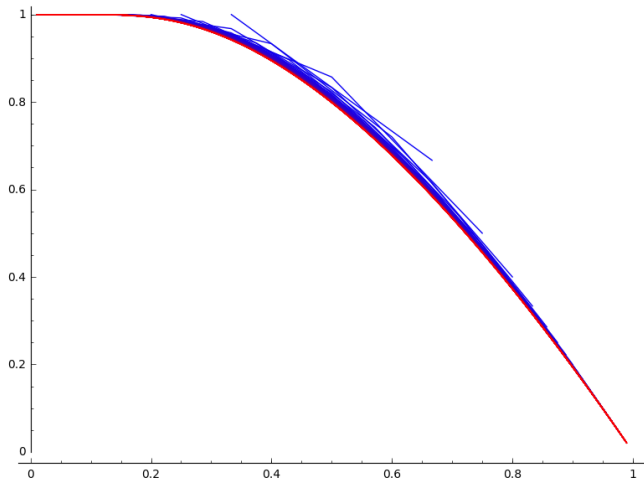
Taylor expansion of  $z = \rho + (z - \rho)$ :

$$\begin{aligned} \left\{ \begin{matrix} n-1 \\ k \end{matrix} \right\} &= \frac{(n-1)!}{k!} \frac{1}{2\pi i} \oint_{|z|=\rho} \phi(z)^k z^{-n-1} \, dz \rho \\ &\quad + \frac{(n-1)!}{k!} \frac{1}{2\pi i} \oint_{|z|=\rho} \phi(z)^k z^{-n-1} (z - \rho) \, dz \\ &= \frac{1}{n} \left\{ \begin{matrix} n \\ k \end{matrix} \right\} \rho + \text{Error} \end{aligned}$$



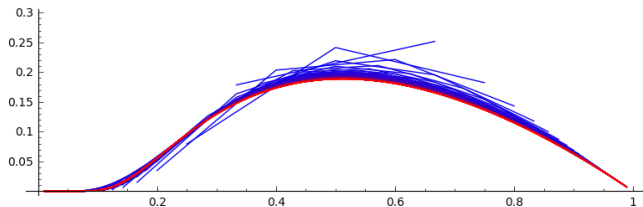
# Direct Approach

$$\frac{k \binom{n-1}{k}}{\binom{n}{k}} \text{ against } \frac{k}{n}, \quad n = 1, \dots, 100$$



# Direct Approach

$$\left( \frac{k \binom{n-1}{k}}{\binom{n}{k}} - \frac{k}{n} \rho \right) n \quad \text{against} \quad \frac{k}{n}, \quad n = 1, \dots, 100$$



**Hard part:** Error analysis to guarantee a uniform error!



# Summary

Cancellations occur:

$$\frac{\text{Large}}{\text{Large}} \sim \text{Small}$$

Bypass cancellations:

$$\int z^{-n-1} \underbrace{\phi(z)^k}_{\text{Large}} \underbrace{f(z)}_{\text{Small}} dz \sim f(\rho) \int z^{-n-1} \phi(z)^k dz$$

$$\frac{\int z^{-n-1} \phi(z)^k f(z) dz}{\int z^{-n-1} \phi(z)^k dz} \sim f(\rho)$$



## Other Examples

- MVUE for the variance of  $\theta$ :

$$\frac{k \binom{n}{k-1}}{\binom{n}{k}} + \left( \frac{\binom{n}{k-1}}{\binom{n}{k}} \right)^2 - \frac{\binom{n}{k-2}}{\binom{n}{k}}$$

- MVUE for sequential capturing until  $r$  marked individuals are caught:

$$k + \frac{\binom{r+k-1}{k-1}}{\binom{r+k-1}{k}}$$

- In network theory
- Extending Markov chains
- ...



## Refining the Asymptotics

$$k \left\{ \begin{matrix} n-1 \\ k \end{matrix} \right\} = \alpha \left\{ \begin{matrix} n \\ k \end{matrix} \right\} f_0(\rho) + \text{Error}$$

where  $f_0(z) = z$  and

$$\text{Error} = \alpha \frac{n!}{k!} \frac{1}{2\pi i} \oint_{|z|=\rho} \phi(z)^k z^{-n-1} (f_0(z) - f_0(\rho)) dz$$



## Refining the Asymptotics

$$k \left\{ \begin{matrix} n-1 \\ k \end{matrix} \right\} = \alpha \left\{ \begin{matrix} n \\ k \end{matrix} \right\} f_0(\rho) + \text{Error}$$

where  $f_0(z) = z$  and

$$\text{Error} = \alpha \frac{n!}{k!} \frac{1}{2\pi i} \oint_{|z|=\rho} \phi(z)^k z^{-n-1} (f_0(z) - f_0(\rho)) dz$$

Apply integration by parts:

$$\text{Error} = \frac{\alpha}{n} \cdot \frac{n!}{k!} \cdot \frac{1}{2\pi i} \int_{|z|=\rho} z^{-n-1} \phi(z)^k f_1(z) dz$$

with

$$f_1(z) = z \frac{d}{dz} \frac{f_0(z) - f_0(\rho)}{\lambda(z) - \lambda(\rho)} \lambda(z), \quad \lambda(z) = \frac{1 - e^{-z}}{z}$$



# Full Asymptotic Expansion

## Theorem (Hwang-K.)

*Uniformly for  $1 \leq k \leq n$*

$$\frac{k \left\{ \begin{matrix} n-1 \\ k \end{matrix} \right\}}{\left\{ \begin{matrix} n \\ k \end{matrix} \right\}} = \alpha f_0(\rho) + \alpha f_1(\rho) \frac{1}{n} + \cdots + \alpha f_{m-1}(\rho) \frac{1}{n^{m-1}} + O(n^{-m})$$

*holds with  $f_0(z) = z$  and*

$$f_{j+1}(z) = z \frac{d}{dz} \frac{f_j(z) - f_j(\rho)}{\lambda(z) - \lambda(\rho)} \lambda(z).$$

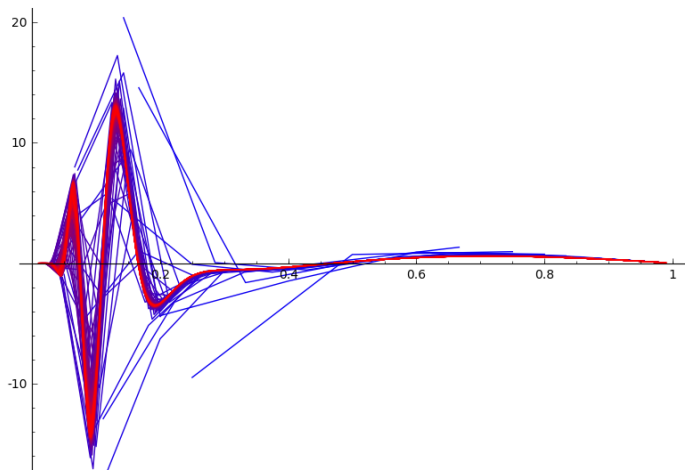
*Under suitable technical conditions on  $\phi$  also applicable to  $[z^n] \phi(z)^k$ .*



# Full Asymptotic Expansion: Stirling Numbers

$$\left( \frac{k \left\{ \begin{smallmatrix} n-1 \\ k \end{smallmatrix} \right\}}{\left\{ \begin{smallmatrix} n \\ k \end{smallmatrix} \right\}} - \alpha\rho - \alpha f_1(\rho) \frac{1}{n} - \cdots - \alpha f_5(\rho) \frac{1}{n^5} \right) n^6 \quad \text{against} \quad \frac{k}{n},$$

$n = 1, \dots, 100$



## Other Examples with Statistical Applications

$$[z^n] \underbrace{\phi(z)^k}_{\text{Large}} \underbrace{f(z)}_{\text{Small}}$$

- B-analogues of  $\{n\}_k$ :  $\phi(z) = e^{2z} - 1$ ,  $f(z) = e^z$
- Binomial coefficient  $\binom{n}{k}$  and Lah numbers:  $\phi(z) = \frac{z}{1-z}$ ,  $f(z) = 1$
- Non-central Stirling numbers of the 2<sup>nd</sup> kind:  $\phi(z) = e^z - 1$ ,  $f(z) = e^{rz}$
- Associative Stirling numbers of the 2<sup>nd</sup> kind:  $\phi(z) = e^z - 1 - z$ ,  $f(z) = 1$
- Many three term recurrences:  $s_{n,k} = a_k s_{n-1,k} + b_k s_{n-1,k-1}$
- ...



# Stirling Numbers of the 1<sup>st</sup> Kind

$\left[ \begin{smallmatrix} n \\ k \end{smallmatrix} \right]$  counts permutations of  $\{1, \dots, n\}$  with  $k$  cycles

$$\left[ \begin{smallmatrix} n \\ k \end{smallmatrix} \right] = \frac{n!}{k!} \frac{1}{2\pi i} \oint_{|z|=\rho} z^{-n-1} \phi(z)^k dz$$

with  $\phi(z) = \log \frac{1}{1-z}$  and

$$\frac{1-\rho}{\rho} \log \frac{1}{1-\rho} = \frac{k}{n}$$





# Stirling Numbers of the 1<sup>st</sup> Kind

$\left[ \begin{smallmatrix} n \\ k \end{smallmatrix} \right]$  counts permutations of  $\{1, \dots, n\}$  with  $k$  cycles

$$\left[ \begin{smallmatrix} n \\ k \end{smallmatrix} \right] = \frac{n!}{k!} \frac{1}{2\pi i} \oint_{|z|=\rho} z^{-n-1} \phi(z)^k dz$$

with  $\phi(z) = \log \frac{1}{1-z}$  and

$$\frac{1-\rho}{\rho} \log \frac{1}{1-\rho} = \frac{k}{n}$$

Formally, we have

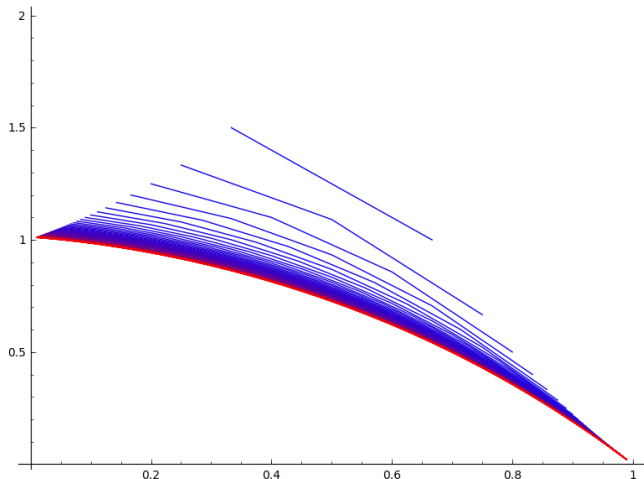
$$\frac{n \left[ \begin{smallmatrix} n-1 \\ k \end{smallmatrix} \right]}{\left[ \begin{smallmatrix} n \\ k \end{smallmatrix} \right]} \sim \rho + f_1(\rho) \frac{1}{n} + \dots$$

for  $1 \leq k \leq n$ .



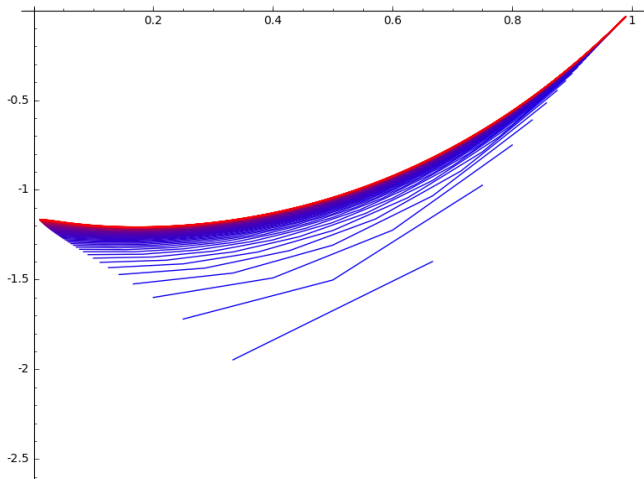
# Stirling Numbers of the 1<sup>st</sup> Kind

$$\frac{n \begin{bmatrix} n-1 \\ k \end{bmatrix}}{\begin{bmatrix} n \\ k \end{bmatrix}} \quad \text{against} \quad \frac{k}{n}, \quad n = 1, \dots, 100$$



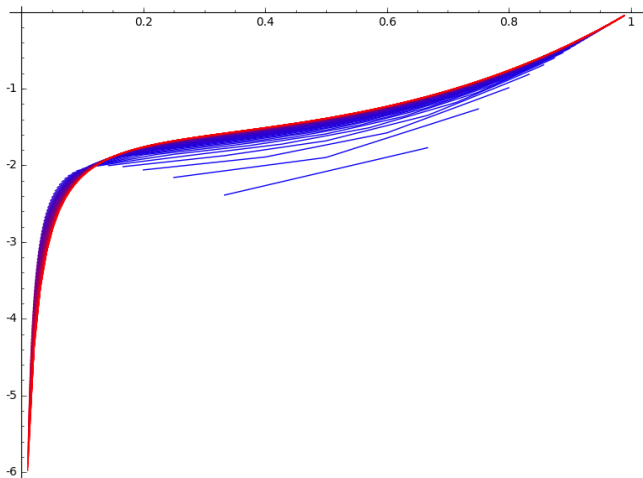
# Stirling Numbers of the 1<sup>st</sup> Kind

$$\left( \frac{n \begin{bmatrix} n-1 \\ k \end{bmatrix}}{\begin{bmatrix} n \\ k \end{bmatrix}} - \rho \right) n \quad \text{against} \quad \frac{k}{n}, \quad n = 1, \dots, 100$$



# Stirling Numbers of the 1<sup>st</sup> Kind

$$\left( \frac{n \begin{bmatrix} n-1 \\ k \end{bmatrix}}{\begin{bmatrix} n \\ k \end{bmatrix}} - \rho - f_1(\rho) \frac{1}{n} \right) n^2 \quad \text{against} \quad \frac{k}{n}, \quad n = 1, \dots, 100$$



# Precomputing the Coefficients

Evaluate at  $z = \rho$

$$f_0(z) = z$$

$$f_{j+1}(z) = z \frac{d}{dz} \frac{f_j(z) - f_j(\rho)}{\lambda(z) - \lambda(\rho)} \lambda(z)$$

by using the rule of de l'Hospital.



# Precomputing the Coefficients

Evaluate at  $z = \rho$

$$f_0(z) = z$$

$$f_{j+1}(z) = z \frac{d}{dz} \frac{f_j(z) - f_j(\rho)}{\lambda(z) - \lambda(\rho)} \lambda(z)$$

by using the rule of de l'Hospital.

Computer said **no**: memory overflow even for  $f_2$ .



# Precomputing the Coefficients

Evaluate at  $z = \rho$

$$f_0(z) = z$$

$$f_{j+1}(z) = z \frac{d}{dz} \frac{f_j(z) - f_j(\rho)}{\lambda(z) - \lambda(\rho)} \lambda(z)$$

by using the rule of de l'Hospital.

Computer said **no**: memory overflow even for  $f_2$ .

$$f_{m+1}(\rho) = f_m^{(1)}(\rho) \rho \frac{d}{dz} \frac{\lambda(z)(z - \rho)}{\lambda(z) - \lambda(\rho)} \Big|_{z=\rho} + \frac{1}{2} f_m^{(2)}(\rho) \rho \frac{\lambda(\rho)}{\lambda'(\rho)}$$

and similar for  $f_m^{(1)}$ ,  $f_m^{(2)}$ , ...

$f_0, \dots, f_{10}$  in less than half an hour (depending on  $\phi$ )

$f_0(z) = z$  and  $f_1$  very fast



# Back to Uniform Sampling of Partitions

Procedure for a uniform partition of  $\{1, \dots, n\}$  with  $k$  parts:

- Precompute  $f_1, \dots$





# Back to Uniform Sampling of Partitions

Procedure for a uniform partition of  $\{1, \dots, n\}$  with  $k$  parts:

- Precompute  $f_1, \dots$
- Repeat:
  - Bernoulli RV  $X_{n,k}$  with

$$\mathbb{P}(X_{n,k} = 1) = \frac{k \binom{n-1}{k}}{\binom{n}{k}} \sim \alpha \rho + \alpha f_1(\rho) \frac{1}{n} + \dots$$

where

$$\frac{1 - e^{-\rho}}{\rho} = \alpha, \quad \text{that is} \quad \rho = \frac{1}{\alpha} + W(-\alpha^{-1} e^{-1/\alpha})$$



# Back to Uniform Sampling of Partitions

Procedure for a uniform partition of  $\{1, \dots, n\}$  with  $k$  parts:

- Precompute  $f_1, \dots$
- Repeat:
  - Bernoulli RV  $X_{n,k}$  with

$$\mathbb{P}(X_{n,k} = 1) = \frac{k \binom{n-1}{k}}{\binom{n}{k}} \sim \alpha \rho + \alpha f_1(\rho) \frac{1}{n} + \dots$$

where

$$\frac{1 - e^{-\rho}}{\rho} = \alpha, \quad \text{that is} \quad \rho = \frac{1}{\alpha} + W(-\alpha^{-1} e^{-1/\alpha})$$

- If  $X_{n,k} = 1$ , then sample a partition of  $\{1, \dots, n-1\}$  with  $k$  parts and add the  $n$ -th element to one
- Otherwise sample a partition of  $\{1, \dots, n-1\}$  with  $k-1$  parts



## What else?

- $X_n$  with probability generating function

$$\frac{[z^n]e^{t\phi(z)}}{[z^n]e^{\phi(z)}}$$

## What else?

- $X_n$  with probability generating function

$$\frac{[z^n]e^{t\phi(z)}}{[z^n]e^{\phi(z)}}$$

- Variance

$$\mathbb{V}X_n = \frac{[z^n](\phi(z)^2 + \phi(z))e^{\phi(z)}}{[z^n]e^{\phi(z)}} - \left( \frac{[z^n]\phi(z)e^{\phi(z)}}{[z^n]e^{\phi(z)}} \right)^2$$



## What else?

- $X_n$  with probability generating function

$$\frac{[z^n]e^{t\phi(z)}}{[z^n]e^{\phi(z)}}$$

- Variance

$$\mathbb{V}X_n = \frac{[z^n](\phi(z)^2 + \phi(z))e^{\phi(z)}}{[z^n]e^{\phi(z)}} - \left( \frac{[z^n]\phi(z)e^{\phi(z)}}{[z^n]e^{\phi(z)}} \right)^2$$

- Cancellation occur
- Same recursive approach works here too



# Variance

For suitable  $\phi$

$$\mathbb{V}X_n \sim \sum_{j \geq 0} h_j(\rho) n^{-j}$$

with explicit, cancellation free expressions for  $h_j(\rho)$ :

$$h_0(\rho) = \phi(\rho)$$

$$h_1(\rho) = \rho \frac{d}{dz} \frac{\phi(z) - \phi(\rho)}{\frac{z\phi'(z)}{n} - 1} + \rho\phi'(\rho) \frac{\phi(z) - \phi(\rho)}{\frac{z\phi'(z)}{n} - 1} \Big|_{z=\rho}$$

with the saddle point

$$\frac{\rho\phi'(\rho)}{n} = 1$$



# Variance

For suitable  $\phi$

$$\mathbb{V}X_n \sim \sum_{j \geq 0} h_j(\rho) n^{-j}$$

with explicit, cancellation free expressions for  $h_j(\rho)$ :

$$h_0(\rho) = \phi(\rho)$$

$$h_1(\rho) = \rho \frac{d}{dz} \frac{\phi(z) - \phi(\rho)}{\frac{z\phi'(z)}{n} - 1} + \rho\phi'(\rho) \frac{\phi(z) - \phi(\rho)}{\frac{z\phi'(z)}{n} - 1} \Big|_{z=\rho}$$

with the saddle point

$$\frac{\rho\phi'(\rho)}{n} = 1$$

For example  $\phi(z) = e^z - 1$ ,  $\phi(z) = (1 - z)^{-a}$ , ...

Again:  $\phi(z) = \log \frac{1}{1-z}$  does not work!



# Conclusion

- Computation of ratios of Stirling numbers
  - precise
  - and efficient
- Precomputation of  $f_j$  in reasonable time (if necessary)
- Also applicable to many other combinatorial sequences
- Formula is uniform, so no knowledge about the relation between  $n$  and  $k$  necessary
- Approach also useful for a direct computation of the variance





# Conclusion

- Computation of ratios of Stirling numbers
  - precise
  - and efficient
- Precomputation of  $f_j$  in reasonable time (if necessary)
- Also applicable to many other combinatorial sequences
- Formula is uniform, so no knowledge about the relation between  $n$  and  $k$  necessary
- Approach also useful for a direct computation of the variance

Thank you for your attention!

