## A Berry-Esseen Theorem for the Lightbulb Process

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## The Lightbulb Process

Motivated by a study in the pharmaceutical industry on the effects of dermal patches designed to activate targeted receptors. An active receptor will become inactive, and an inactive one active, if it receives a dose of medicine released from the dermal patch.

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In the lightbulb process of Rao, Rao and Zhang, on days $r=1, \ldots, n$, out of $n$ lightbulbs, all initially off, exactly $r$ bulbs, selected uniformly and independent of the past, have their status changed from off to on, or vice versa.

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Their histograms indicated that $X$ is asymptotically normal, but left the question open. 'Out of the box' central limit theorems seem to not apply easily to this process.

## Mean and Variance

For $s=1, \ldots, n$, let
$\lambda_{n, 1, s}=1-\frac{2 s}{n} \quad$ and $\quad \lambda_{n, 2, s}=1-\frac{4 s}{n}+\frac{4 s(s-1)}{n(n-1)}$,
and $\lambda_{n, b, \mathbf{s}}=\prod_{r=1}^{n} \lambda_{n, b, s_{r}}$. Then with switch pattern $\mathbf{s}=\left(s_{1}, \ldots, s_{n}\right)$,

$$
E X_{\mathbf{s}}=\frac{n}{2}\left(1-\lambda_{n, 1, \mathbf{s}}\right)
$$

and

$$
\operatorname{Var}\left(X_{\mathbf{s}}\right)=\frac{n}{4}\left(1-\lambda_{n, 2, \mathbf{s}}\right)+\frac{n^{2}}{4}\left(\lambda_{n, 2, \mathbf{s}}-\lambda_{n, 1, \mathbf{s}}^{2}\right) .
$$

## Berry Esseen Theorem for Lightbulb

Theorem 1 Let $X$ be the number of bulbs on at the terminal time $n$, an even integer, and $\mu=n / 2$ and $\sigma^{2}=\operatorname{Var}(X)$. Then
$\sup _{z \in \mathbb{R}}\left|P\left(\frac{X-\mu}{\sigma} \leq z\right)-P(Z \leq z)\right| \leq \frac{n}{2 \sigma^{2}} \Psi_{0}+1.64 \frac{n}{\sigma^{3}}+\frac{2}{\sigma}$
where $Z$ is standard normal and

$$
\Psi_{0} \leq \frac{1}{2 \sqrt{n}}+\frac{1}{2 n}+e^{-n / 2} \quad \text { for } n \geq 6
$$

Yields a bound of order $O\left(n^{-1 / 2}\right)$ as $n \rightarrow \infty$.

## Composition Markov chains of multinomial

## type

Considered by Zhou and Lange, such chains in general are based on a $d \times d$ Markov transition matrix $P$ which describes the transition of a single particle in a system of $n$ identical particles, a subset of which is selected uniformly to undergo transition at each time step according to $P$.

Very complete spectral decompositions of such chains may be available; this decomposition becomes essential to the calculation of the Berry Esseen bound for this problem.

Ehrenfest chains, Hoare-Rahman chains, Kimura's model for DNA base-pair substitution.

## Size Bias Coupling

Let $X$ be nonnegative with finite mean $\mu$. Recall $X^{s}$ has the $X$-sized biased distribution if

$$
E[X f(X)]=\mu E\left[f\left(X^{s}\right)\right]
$$

for all smooth $f$.
If $X=\sum_{i=1}^{n} X_{i}$ is the sum of exchangeable indicators, then one may form $X^{s}$ by picking $i$ uniformly and summing the variables having distribution

$$
\mathcal{L}\left(X_{1}^{i}, \ldots, X_{n}^{i}\right)=\mathcal{L}\left(X_{1}, \ldots, X_{n} \mid X_{i}=1\right)
$$

## Size biased Coupling

With $W=(X-\mu) / \sigma$, and $W^{s}=\left(X^{s}-\mu\right) / \sigma$,

$$
\begin{aligned}
& E(h(W)-E h(Z)) \\
&= E\left(f^{\prime}(W)-W f(W)\right) \\
&= E\left(f^{\prime}(W)-\frac{\mu}{\sigma}\left(f\left(W^{s}\right)-f(W)\right)\right) \\
&= E\left(f^{\prime}(W)\left(1-\frac{\mu}{\sigma}\left(W^{s}-W\right)\right)\right. \\
&\left.-\frac{\mu}{\sigma} \int_{0}^{W^{s}-W}\left(f^{\prime}(W+t)-f^{\prime}(W)\right) d t\right) .
\end{aligned}
$$

When monotone $W^{s} \geq W$.

## Concentration Inequality

Size bias version of exchangeable pair concentration inequality of Shao and Su (2005).

Lemma 1 Let $X$ be a nonnegative random variable with mean $\mu$ and variance $\sigma^{2}$, both finite and positive, and let $X^{s}$ be given on the same space as $X$, having the $X$ size biased distribution and satisfying $X^{s} \geq X$. Then with $W=(X-\mu) / \sigma$ and $W^{s}=\left(X^{s}-\mu\right) / \sigma$, for any $z \in \mathbb{R}$ and $a \geq 0$,

$$
\frac{\mu}{\sigma} E\left(W^{s}-W\right) \mathbf{1}_{\left\{W^{s}-W \leq a\right\}} \mathbf{1}_{\{z \leq W \leq z+a\}} \leq a .
$$

## Monotone Size Biased Coupled Pair

Theorem 2 Let $X$ be a nonnegative random variable with mean $\mu$ and variance $\sigma^{2}$, both finite and positive, and let $X^{s}$ be given on the same space as $X$, with the $X$ size biased distribution, satisfying $X \leq X^{s} \leq X+B$ for some $B>0$. Then with $W=(X-\mu) / \sigma$, we have

$$
\sup _{z \in \mathbb{R}}|P(W \leq z)-P(Z \leq z)| \leq \frac{\mu}{\sigma^{2}} \Psi+0.82 \frac{\delta^{2} \mu}{\sigma}+\delta
$$

where

$$
\Psi=\sqrt{\operatorname{Var}\left(E\left(X^{s}-X \mid X\right)\right)} \quad \text { and } \quad \delta=B / \sigma .
$$

## Size biased coupling: Lightbulb

The number of bulbs on at time $n$ is

$$
X=\sum_{k=1}^{n} X_{k}
$$

where $X_{k}$ is the indicator that bulb $k$ is on. Need to select bulb at random, and turn it on if not already, keeping correct conditional distribution, and without greatly affecting the status of the other bulbs.

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With $X_{r k}$ the switch variable of bulb $k$ at time $r$, we have

$$
X_{k}=\left(\sum_{r=0}^{n} X_{r k}\right) \bmod 2
$$

## Size biased coupling: Lightbulb

Let $n=2 m$ and $\mathbf{X}=\left\{X_{r k}, r, k=1, \ldots, n\right\}$. For given bulb $i$, if $X_{i}=1$ then set $\mathbf{X}^{i}=\mathbf{X}$.

Otherwise $J^{i} \sim \mathcal{U}\left\{j: X_{n / 2, j}=1-X_{n / 2, i}\right\}$, independent of $\left\{X_{r k}: r \neq n / 2, k=1, \ldots, n\right\}$. Let $\mathbf{X}^{i}$ have components

$$
X_{r k}^{i}=\left\{\begin{array}{cl}
X_{r k} & r \neq n / 2 \\
X_{n / 2, k} & r=n / 2, k \notin\left\{i, J^{i}\right\} \\
X_{n / 2, J^{i}} & r=n / 2, k=i \\
X_{n / 2, i} & r=n / 2, k=J^{i} .
\end{array}\right.
$$

## Size biased coupling: Lightbulb

In words, if $X_{i}=0$ then uniformly select a $j$ whose switch variable $X_{n / 2, j}$ at stage $n / 2$ is opposite to $X_{n / 2, i}$, and interchange.

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If $I$ is the uniformly chosen index, and $J$ the index of the bulb with opposite status at stage $n / 2$, then

$$
X^{s}-X=21_{\left\{X_{I}=0, X_{J}=0\right\}} .
$$

The coupling is monotone and bounded.

## Conditional Variance Calculation, $n=2 m$

With $\mathcal{F}$ the $\sigma$-algebra generated by the switch variables $\mathbf{X}$,

$$
E\left(X^{s}-X \mid \mathcal{F}\right)=\frac{4}{n^{2}} \sum_{i \neq j} \mathbf{1}_{\left\{X_{i}=0, X_{j}=0, X_{n / 2, i}=0, X_{n / 2, j}=1\right\}} .
$$

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The variance calculation requires the computation of (four) joint probabilities such as $g_{2,2, n,(1, \ldots, n), n / 2},\left(g_{\alpha, \beta, n, \mathbf{s}, l}\right)$

$$
\begin{aligned}
& P\left(X_{i_{1}}=0, X_{j_{1}}=0, X_{i_{2}}=0, X_{j_{2}}=0\right. \\
& \left.X_{n / 2, i_{1}}=0, X_{n / 2, j_{1}}=1, X_{n / 2, i_{2}}=0, X_{n / 2, j_{2}}=1\right)
\end{aligned}
$$

Conditioning on switches at stage $n / 2$ they later become 'initial conditions.'

## Spectral Decomposition

For a given subset of $b$ of the $n$ bulbs, the eigenvalues of one, and multiple steps of the chain are given by
$\lambda_{n, b, s}=\sum_{t=0}^{b}\binom{b}{t}(-2)^{t} \frac{(s)_{t}}{(n)_{t}} \quad$ and $\quad \lambda_{n, b, \mathbf{s}}=\prod_{r=1}^{k} \lambda_{n, b, s_{r}}$,
where $(n)_{k}=n(n-1) \cdots(n-k+1)$ denotes the falling factorial, and the empty product is 1 .

## Spectrum of Composition Markov Chains of Multinomial Type

In the case of the lightbulb chain there are $d=2$ states and the transition matrix $P$ of a single bulb is given by

$$
P=\left[\begin{array}{ll}
0 & 1 \\
1 & 0
\end{array}\right]
$$

With $b \in\{0,1, \ldots, n\}$ let $P_{n, b, s}$ be the $2^{b} \times 2^{b}$ transition matrix of a subset of size $b$ of the $n$ total lightbulbs when $s$ of the $n$ bulbs are selected uniformly to be switched. Letting $P_{n, 0, s}=1$ for all $n$ and $s$, and $I_{2}$ the $2 \times 2$ identity matrix, for $n \geq 1$ the matrix $P_{n, b, s}$ is given recursively by $P_{n, b, s}=\frac{s}{n}\left(P \otimes P_{n-1, b-1, s-1}\right)+\left(1-\frac{s}{n}\right)\left(I_{2} \otimes P_{n-1, b-1, s}\right)$.

## Spectral Decomposition

Transition matrices are simultaneously diagonalizable by

$$
P_{n, b, s}=\otimes^{b} T^{-1} \Gamma_{n, b, s} \otimes^{b} T,
$$

where $\Gamma_{n, b, s}=\operatorname{diag}\left(\lambda_{n, a_{1}, s}, \ldots, \lambda_{n, a_{2^{b}, s}}\right)$, e.g.,

$$
\mathbf{a}_{1}=(0,1), \quad \mathbf{a}_{2}=(0,1,1,2) \quad \text { and } \quad \mathbf{a}_{3}=(0,1,1,2,1,2,2,3),
$$

and

$$
T=\frac{1}{\sqrt{2}}\left[\begin{array}{rr}
1 & 1 \\
-1 & 1
\end{array}\right] .
$$

## Spectral Decomposition

For calculation of $g_{2,2, n,(1, \ldots, n), n / 2}$,

$$
\begin{aligned}
& P\left(X_{i_{1}}=0, X_{j_{1}}=0, X_{i_{2}}=0, X_{j_{2}}=0\right. \\
& \left.X_{n / 2, i_{1}}=0, X_{n / 2, j_{1}}=1, X_{n / 2, i_{2}}=0, X_{n / 2, j_{2}}=1\right)
\end{aligned}
$$

spectral decomposition yields

$$
g_{2,2, n, \mathbf{s}, l}=\frac{1}{16}\left(1-2 \lambda_{n, 2, \mathbf{s}_{l}}+\lambda_{n, 4, \mathbf{s}_{l}}\right) \frac{\left(s_{l}\right)_{2}\left(n-s_{l}\right)_{2}}{(n)_{4}}
$$

where

$$
\mathbf{s}_{l}=\left(s_{1}, \ldots, s_{l-1}, s_{l+1}, \ldots, s_{n}\right)
$$

## Spectral Decomposition

Now turning to $\lambda_{n, 4, \mathbf{s}}$, for $n \geq 4$ consider the fourth degree polynomial
$f_{4}(x)=1-\frac{8 x}{n}+\frac{24(x)_{2}}{(n)_{2}}-\frac{32(x)_{3}}{(n)_{3}}+\frac{16(x)_{4}}{(n)_{4}}, \quad 0 \leq x \leq n$.
It can be checked that the four roots of $f_{4}(x)$ are given by

$$
x_{1 \pm}=\frac{n \pm \sqrt{\sqrt{2} \sqrt{3 n^{2}-9 n+8}+3 n-4}}{2}
$$

with $0<x_{1-}<x_{2-}<x_{2+}<x_{1+}<n$ where

$$
x_{2 \pm}=\frac{n \pm \sqrt{-\sqrt{2} \sqrt{3 n^{2}-9 n+8}+3 n-4}}{2}
$$

$$
f_{4}(x)
$$



## Need Bounds to compute Conditional Variance

$$
\left|f_{4}(x)\right| \leq\left\{\begin{array}{cl}
\frac{6}{(n-3)^{2}} & \text { for } x \in\left[x_{1-}, x_{1+}\right] \\
f_{2}^{2}(x) & \text { for } x \notin\left[x_{1-}, x_{1+}\right], x \in[0, n]
\end{array}\right.
$$

where

$$
f_{2}(x)=1-\frac{4 x}{n}+\frac{4(x)_{2}}{(n)_{2}}, \quad 0 \leq x \leq n .
$$

$$
\lambda_{n, 4, \mathrm{~s}}
$$

$$
\left|\lambda_{n, 4, \mathbf{s}}\right|=\prod_{s=0}^{\left\lfloor x_{1-}\right\rfloor}\left|\lambda_{n, 4, s}\right| \prod_{s \in \mathbf{u}}\left|\lambda_{n, 4, s}\right| \prod_{s=\left\lceil x_{1+}\right\rceil}^{n}\left|\lambda_{n, 4, s}\right|
$$

Obtain

$$
\left|\lambda_{n, 4, \mathbf{s}}\right| \leq \frac{1}{2} e^{-n} \quad \text { for } n \geq 6
$$

for $s=(1,2, \ldots, n / 2-1, n / 2+1, \ldots, n)$.

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Yields a bound of order $O\left(n^{-1 / 2}\right)$ as $n \rightarrow \infty$.

## Stein's Method for the Lightbulb Process

I hope that the talk was clear and illuminating


## To Louis:

Thank you for all your hard work, and Happy Birthday

