





• Consider the Anytime-Valid T-Test likelihood ratio:

$$
S^{(n)} = \frac{\int f_{\delta_1 \sigma, \sigma}(X^n) \left(\frac{1}{\sigma}\right) d\sigma}{\int f_{0, \sigma}(X^n) \left(\frac{1}{\sigma}\right) d\sigma} = \frac{g_{\delta_1}(V^n)}{g_0(V^n)}
$$

- This is a test martingale, hence an e-process under  $H_0 = \{F_{0,\sigma}: \sigma > 0\}$
- [Perez-Ortiz et al.  $Annals$ , 2024] For fixed  $n$ , it is also an e-variable under larger null

 $H'_0 = \{F_{\mu,\sigma}: \sigma > 0, \mu \leq 0\}$  (relevant in 1-sided testing)

- Aaditya and Hongian asked: is  $S^{(n)}$  also an e-process under  $H'_0$  ?
- **…embarrasingly, we don't know!!!**

## **A Paper You Must Read . (Period.)**

- Variation Diminishing Transformations: A Direct Approach to Total Positivity and Its Statistical Applications. L. Brown, I. Johnstone and K. B. MacGibbon, *JASA* 1981
- Don't get scared by the title, which with hindsight is most ill-chosen!
- Paper makes **Karlin**'s theory of the 1950/1960s about monotone likelihood ratios, stochastic dominance etc. much more accessible by **completely avoiding** the highly involved concept of 'total positivity', which is central in Karlin's treatment



Samuel Karlin

### **Sign Changes**

- Let  $X \subset \mathbb{R}$  be an interval and  $f: X \to \mathbb{R}$  be a continuous function.
- $S^{-}(f)$  stands for the number of sign changes of f
- Generalization to f defined on finite  $X \subset \mathbb{R}$  : let  $x'_1 \leq \cdots \leq x'_n$  be the ordered sequence of elements of X. Then  $S^{-}(f)$  is the number of sign changes of  $(f(x'_1),..., f(x'_n))$ , not counting 0s.

### **VR: Variation Reducing**

Let  $X \subset \mathbb{R}$ ,  $\Theta \subset \mathbb{R}$  and  $f: \Theta \times X \to \mathbb{R}^+_0$ (write  $f_{\theta}(x)$  and think of it as density of X under some measure  $F_{\theta}$ ) We say "f is  $VR_{n+1}$  on X with parameter  $\Theta$ " if for all nonnegative measures v on X and functions  $q: X \to \mathbb{R}$  with  $\int |q|d \nu > 0$ , we have:

 $S^-(q) \leq n \Rightarrow S^-(\gamma) \leq S^-(q)$ 

where  $\gamma(\theta) \coloneqq \int f_{\theta}(x) g(x) \nu(dx)$ .

If  $f_{\theta}$  is indeed a probability measure relative to v this means: if the function g changes sign  $m \leq n$  times when we vary x, then its expectation changes sign at most  $m$  times when we vary  $\theta$ 

# **Example:**, **Exponential Families are**

•  $f_{\theta}(x) =$  $\exp(\theta \cdot x)$  $\frac{\exp(\theta \cdot x)}{\int \exp(\theta \cdot x) d\rho(x)}$  with  $\Theta$  the natural parameter space defines an (arbitrary) 1-dim exp family;  $P_{\theta}$  has density  $f_{\theta}$  relative to  $\rho$ 

- It can be shown that any such f is  $VR<sub>2</sub>$ . This immediately implies that for any monotone increasing function g, we have that  $E_{P_{\theta}}[g(X)]$  is an monotone increasing function in  $\theta$
- ...but this property is also known as stochastic dominance!
- …and it is well-known that for 1-dim exp families,  $P_{\theta}$  stochastically dominates  $P_{\theta}$  whenever  $\theta > \theta'$

### **is stochastic dominance!**

- For many other families besides exponential families, it can also be shown that they are  $VR<sub>2</sub>$ . For example, the noncentral t- and  $\chi^2$ -families are  $VR<sub>2</sub>$  as well.
- In fact stochastic dominance is equivalent to  $VR<sub>2</sub>$

### $VR_{\infty}$

• 1-dim exponential families, noncentral  $\chi^2$  and noncentral F families are even  $VR_{\infty}$  , which abbreviates " $VR_{n+1}$  for all  $n$ "

- Taking expectation under  $\theta$  of any function which changes sign  $n$  times and varying  $\theta$  gives you a function that changes sign at most n times
- There are many more families with finite or infinite  $VR$  properties.
- **But how to prove this for any given family?**

### **Central Theorem (Deep)**

• Theorem 3.1 (going back to Karlin's works of the 1950s and 1960s):

f is  $VR_{n+1}$  on X with parameter  $\Theta$  if and only if

for every pair of finite subsets  $X' \subset X$ ,  $\Theta' \subset \Theta$ , both with  $n + 1$  elements, f is  $VR_{n+1}$  on X' with parameter  $\Theta'$ 

The second, "finite" form is often quite easy to check: involves only summation, no need to perform integrals relative to all measures on interval  $X$ !

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$$

• [Perez-Ortiz et al. *Annals*, 2024] For fixed n, it is an e-variable under larger null

$$
H_0' = \{ F_{\delta \sigma, \sigma} : \sigma > 0, \delta \le 0 \}
$$

This null sometimes more relevant (one-sided testing)

- Aaditya and Hongian asked: is  $S^{(n)}$  also an e-process under  $H'_0$  ?
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Proposition:  $S^{(n)}$  is e-variable, i.e.  $\forall \delta \leq 0$ :  $u(\delta) \leq 1$  with  $u(\delta) \coloneq \mathbf{E}_{H_{\delta}}[S^{(n)}]$ 

- *VR*-Reformulation of proof:  $u(0) = 1$  (trivially by cancellation)  $\bullet$
- LR is monotone increasing in  $T_n$ . Since 1-parameter family  $h_{\delta}(T_n)$  is  $\bullet$  $VR<sub>2</sub>$ , so is  $u(\delta)$ . The result follows!

$$
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$$

To prove  $(S^{(n)})$  $\overline{n}$ is e-process, sufficient to prove:

$$
\forall n, v^n, \delta \le 0: \ u(\delta) | v^n \le 1 \text{ with } u(\delta | v^n) \coloneqq \mathbf{E}_{H_\delta} \left[ \frac{h_{\delta_1}(T_n | v^{n-1})}{h_0(T_n | v^{n-1})} | V^{n-1} = v^{n-1} \right]
$$

Conjecture: for any constant  $C$ , (a) conditional LR inside expectation has at most 1 extremum, (b) conditional densities are  $VR<sub>3</sub>$ . If (a)+(b) true, then  $u(\delta | v^n)$  has at most 1 extremum, and the result seems provable