Significance In the On-Off Problem

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What Can HEP and Astrophysics Practice Teach Each Other? Astrophysics (especially  $\gamma$  ray) aims at simple formulae (very fast) calculates  $\sigma$ 's directly (Asymptotic Normal) hope it's a good formula HEP (especially Fermilab practice) calculates probabilities by MC (general; slow) translates into  $\sigma$ 's for communication loses track of analytic structure

Cousins, Linnemann, Tucker, NIM A 595 (2008) 480-501

# Observed vs. Prospective Significance

- This discussion: Observed Significance (of my data)
  - Post-hoc: (after data)
  - Need definition of Z
  - Choice of Zmin for observational claim

= max P(observed|background)

- Prospective Observability (before data, to optimize expt.)
  - Should consider  $\underline{Pr}$  ( Z > Zmin ) (making observational claim)

### Backgrounds in Astro and HEP

Astrophysics: On-source vs. Off-source
 – side observation with τ = Toff/Ton (sensitivity ratio)

$$\hat{\mu}_b = n_{off} / \tau;$$

 HEP: estimate background in defined sideband region
 τ is ratio of signal and sideband region

## Li and Ma (Gamma Ray)

$$Z_{PL} = \sqrt{2} \sqrt{x \bullet Ln[(1+\tau)(\frac{x}{x+y})]} + y \bullet Ln[(\frac{1+\tau}{\tau})(\frac{y}{x+y})]$$
$$x = Non; \quad y = Noff$$

Generic test for composite hypothesis + Wilks' Theorem (conditions not satisfied)

#### **Binomial Proportion Test: Ratio of Poisson Means**

P-value = Pr Binomial(  $\geq$ Non | p, k) where  $p = \alpha/(1+\alpha)$  $p - value = \sum_{j=x}^{k} {k \choose j} p^{j} (1-p)^{k-j}$ 

Holds k = Non + Noff fixed (k a nuisance parameter)

UMPU (Uniformly Most Powerful Unbiased)

for **Composite Hypothesis test**  $\mu_{on} / \alpha \mu_{off} > 1$ 

**Optimal?** Not continuous—issues for small n

Not in common use; probably should be

Known in HEP and Astrophysics: but not as optimal, nor standard procedure

- Zhang and Ramsden claim too conservative for Z small Even if true, we want Z > 4
- Closed form in term of special functions, or sums
  - Applying for large N requires some delicacy; **ZPL**

## **Bayesian Methods**

- In common use in HEP
  - Cousins & Highland "smeared likelihood" efficiency
- Predictive Posterior (after background measurement)
   P(Non | Noff) (integrate posterior μ<sub>b</sub>)
   A flat prior for background, gives Gamma dist. for p(μ<sub>b</sub> | Noff)
  - P value calc using Gamma:(also Alexandreas--Astro)IDENTICAL to Frequentist Binomial Test

Predictive Posterior Bayes P-value (HEP)

In words: tail sum averaged over Bayes posterior for mean

or: integrate before sum

$$P-value(x, y) = \sum_{j=x}^{\infty} p(j|y)$$

$$p(j \mid y) = \int p(j \mid \mu) p(\mu \mid y) d\mu$$
$$p(j \mid \mu) = \frac{\mu^{j} e^{-\mu}}{j!}$$

$$p_{\Gamma}(\mu \mid y) = \frac{\beta^{y} e^{-\beta}}{y!}, \quad \beta = \mu / \alpha$$

 $p_N(\mu \mid y) = Normal \left[ (\mu - b) / \delta b \right]$ 

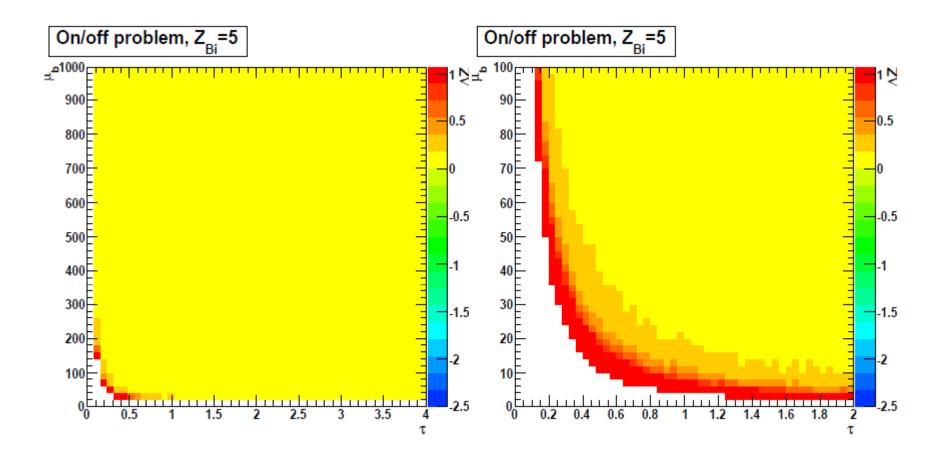
## Comparing the Methods

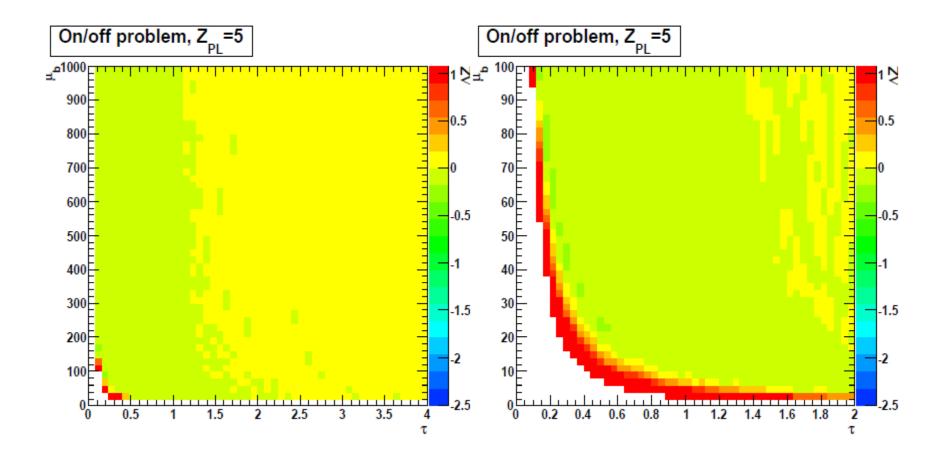
Some test cases from published literature And a few artificial cases Range of Non, Noff values Different  $\tau$  values (mostly > 1)

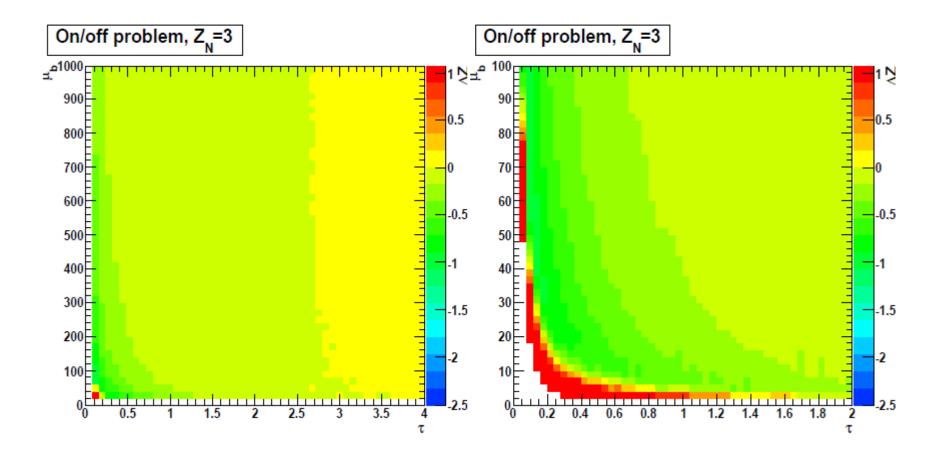
Can show some approximate Z's strictly > others including popular shortcut formulas
Others cross over as τ varies

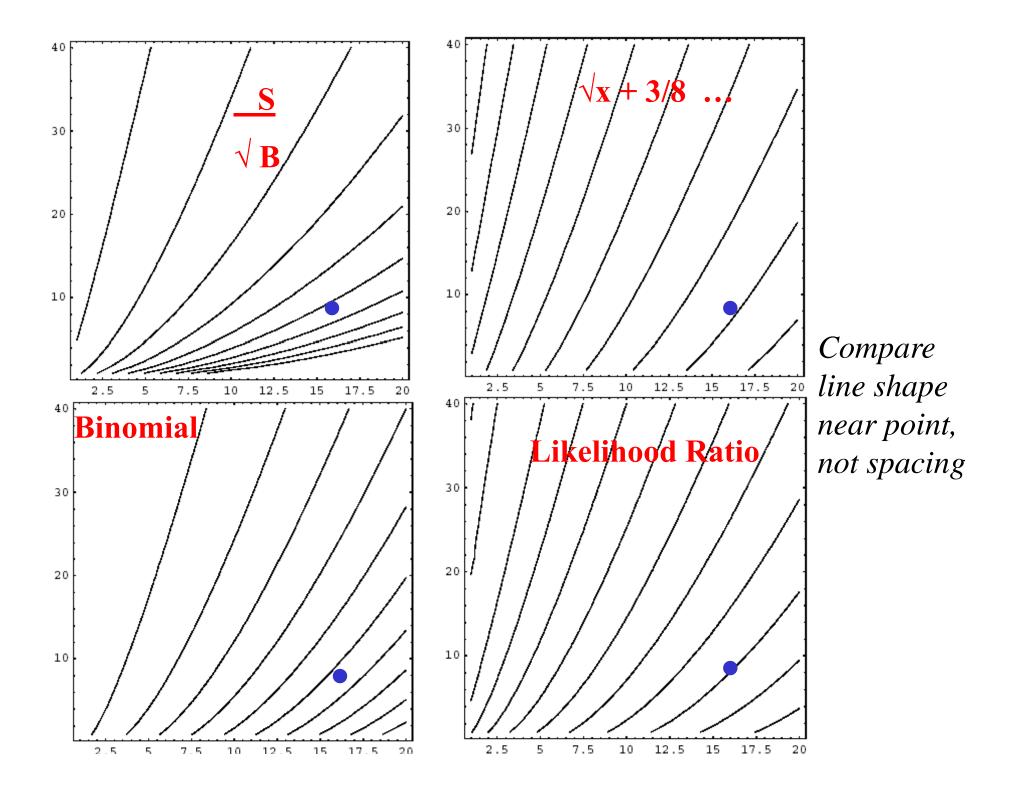
Coverage Calculations (Tucker, Cousins)

Reference	[40]	[41]	[42]	[43]	[44]	[44]	[45]	[46]
$n_{on}$ $n_{off}$ au $\hat{\mu}_b$ $s = n_{on} - \hat{\mu}_b$ $\sigma_b$ $f = \sigma_b / \hat{\mu}_b$ Reported $p$ Reported $Z$	4 5 5.0 1.0 3.0 0.447 0.447	6 18.78 14.44 <b>1.3</b> 4.7 <b>0.3</b> 0.231 <b>0.003</b> 2.7	9 17.83 4.69 <b>3.8</b> 5.2 <b>0.9</b> 0.237 <b>0.027</b> 1.9	17 40.11 10.56 3.8 13.2 0.6 0.158 2E-06 4.6	50 55 2.0 27.5 22.5 3.71 0.135	67 15 0.5 30.0 37 7.75 0.258	200 10 0.1 100.0 100 31.6 0.316	523 2327 5.99 388.6 134 8.1 0.0207 5.9
See conclusion $Z_{\text{Bi}} = Z_{\Gamma}$ binomial $Z_{\text{N}}$ Bayes Gaussian $Z_{\text{PL}}$ profile likelihood $Z_{\text{ZR}}$ variance stabilization	<b>1.66</b> 1.88 1.95 1.93	<b>2.63</b> 2.71 2.81 2.66	<b>1.82</b> 1.94 1.99 1.98	<b>4.46</b> 4.55 4.57 4.22	<b>2.93</b> 3.08 3.02 3.00	2.89 3.44 3.04 3.07	<b>2.20</b> 2.90 2.38 2.39	5.93 5.93 5.93 5.86
Not recommended $Z_{\text{BiN}} = s/\sqrt{n_{\text{tot}}/\tau}$ $Z_{\text{nn}} = s/\sqrt{n_{\text{on}} + n_{\text{off}}/\tau^2}$ $Z_{\text{ssb}} = s/\sqrt{\mu_{\text{b}} + s}$ $Z_{\text{bo}} = s/\sqrt{n_{\text{off}}(1 + \tau)/\tau^2}$	2.24 1.46 1.50 2.74	3.59 1.90 1.92 3.99	2.17 1.66 1.73 2.42	5.67 3.17 3.20 6.47	3.11 2.82 3.18 3.50	<b>2.89</b> 3.28 4.52 3.90	<b>2.18</b> 2.89 7.07 3.02	6.16 5.54 5.88 6.31
Ignore $\sigma_b$ $Z_P$ Poisson: ignore $\sigma_b$ $Z_{sb} = s/\sqrt{\hat{\mu}_b}$	2.08 3.00	2.84 4.12	2.14 2.67	4.87 6.77	3.80 4.29	5.76 6.76	8.76 10.00	6.44 6.82
Unsuccessful ad hockery Poisson: $\mu_{\rm b} \rightarrow \hat{\mu}_{\rm b} + \sigma_{\rm b}$ $s/\sqrt{\hat{\mu}_{\rm b} + \sigma_{\rm b}}$	1.56 2.49	2.51 3.72	1.64 2.40	<b>4.47</b> 6.29	3.04 4.03	4.24 6.02	5.51 8.72	6.01 6.75









## What did we learn?

- Shapes of tails matter at 3-5 sigma ZBi: no undercoverage; can overcover for small N Recommended
- ZPL quite reasonable behavior (despite Wilks); pretty fast to calculate ZN undercovers, worse at high Z (Cranmer) S/√B and S/√(S+B): just don't: tails are wrong Small τ is hard to get right

## Summary

#### Should use Binomial Test for small N, large Zmin

Good Frequentist Properties
smallest N, overcovers a bit
numerically, more work than ZPL
Binomial Test and L. Ratio have roots in Hyp Testing

For high and moderate N, ZPL Likelihood Ratio Good Not so much for low N or negative

Most wrong formulae overestimate significance  $S/\sqrt{B}$  is way too optimistic—ignores uncertainty in B You MUST check properties ZN has coverage problems at large Zmin

#### References

Li & Ma Astroph. Journ. 272 (1983) 314-324 Zhang & Ramsden Experimental Astronomy 1 (1990) 145-163 Fraser Journ. Am. Stat. Soc. 86 (1990) 258-265 Alexandreas et. al. Nuc. Inst. & Meth. A328 (1993) 570-577

Gelman et. al., Bayesian Data Analysis, Chapman & Hall (1998) (predictive p-value terminology)