

$\frac{i}{n+1}$ only for: p-p ($F_n(x), F^*(x)$)
percentile
"Smoothed"

"Max covered loss" refer to X

Binomial-Beta

$N \sim \text{Bin}(m, q)$

$Q \sim \text{Beta}\left(\begin{smallmatrix} a \\ b \end{smallmatrix}\right)$ $\times$ claims

$Q \sim \text{Beta}\left(\begin{smallmatrix} a^* = a + x \\ b^* = b + m - x \\ \theta \end{smallmatrix}\right)$

3 Points for Emp Bayes

- ① non-par have ugly $\hat{v}$
- ② semipar have $\hat{v} = \bar{X}$
- ③ semipar $\wedge$ uni have $a^2 = s^2 - v^2$

$g(x) = a^x$
 $g'(x) = \ln a \cdot a^x$

$\text{Var}_p(X)$ rd ↑

3 Points for Bayes/Bühl Lines

- ① Avg. the line
- ② Linear Bühl
- ③ Bayes has realistic bounds

$a, b = nq \pm s \pm \sqrt{npq}$
rounds: ↑ ↓

Cart prch geo
w/ Mom
 $\sigma^2 \geq \bar{X}$

Var(MLE)

Asymp. $\Rightarrow$ Info

$I(\theta) = -E[l''(\theta)]$

Standard $\Rightarrow$ Delta Method

$\text{Var}(g(X)) = \Delta_{\theta(X)}^2 \text{Var}(X)$

or $\hat{\Sigma} = g'(\bar{X}) \rightarrow \text{differential } \bar{X}^{-1}$
to $-\bar{X}^{-2}$

Var $(X-d)_+$ of $X \sim \exp(\theta)$

$(X-d)_+ = \begin{cases} 0 & X < d \\ X-d & X > d \end{cases}$

$\text{Var}((X-d)_+) = E[\text{Var}(Y|\theta)] + \text{Var}(E[Y|\theta])$
 $= \theta^2 P(X > d) + (\theta - 0)^2 P(X < d) P(X > d)$
↑
(memoryless & conditioned on being $> d$)

Log-Likelihood

$z(l_F - l_0) \sim \chi^2_{df \text{ of } \theta_{new}}$
"King of the hill, adding param"

Bühl = Bayes

For any conj priors

"Like χ^2 " - no adj

"Smirnov-Darling is needy" - needs adj

"Darling is a slut" $\rightarrow n \rightarrow$ irrelevant
&
heavy tail weight

$TVAR_p = e_L(\pi_p) + \pi_p$

$= \frac{E[L] - E[L \wedge \pi_p]}{1-p} + \pi_p$

$TVAR_p(aX-b) = a TVAR_p(X) - b$

$P(z) = E[z^x]$ $P'(0) = p_1$
 $P'(1) = E[X]$

Bootstrap
 $MSE = E_{emp}(\hat{\theta}_i - \theta_{emp})$
varies ↑ constant ↑

Y^P is already
continuous!

Kurtosis $\frac{E[(X-\mu)^4]}{\sigma^4} \rightarrow$ expand it

$E[X \wedge d]$ is the welfare
 $= \int_0^d x f(x) dx + d P(X > d)$
 (or $= \int_0^d S(x) dx$ as all other do)

Consistent if ① bias $\rightarrow 0$
 ② Var $\rightarrow 0$

Empirical Unspecified Ogive $A \rightarrow F_n(x)$
 $\frac{i}{n}$ | $\frac{i}{n+1}$ Smoothed Percentile Method
 p-p plot $(F_n(x), F^*(x))$
 $DGS = F_n(x) - F^*(x)$ (split integrals)

Fit MoM to dist
 ↳ no geo (other 3 ok)

Var (MLE)

Asymptotic: $I_n F_0$

2nd Derivs

$$I(\theta) = - \begin{pmatrix} E \left[\frac{d^2}{d\theta_1^2} \ell(\theta) \right] & E \left[\frac{d^2}{d\theta_1 d\theta_2} \ell(\theta) \right] \\ E \left[\frac{d^2}{d\theta_2 d\theta_1} \ell(\theta) \right] & E \left[\frac{d^2}{d\theta_2^2} \ell(\theta) \right] \end{pmatrix}$$

$$Var(\theta) = I^{-1}(\theta)$$

$$I(\theta) = -E[\ell''(\theta)]$$

Standard: Delta Method

1st Derivs

$$Var(g(\hat{x})) = [\Delta_\theta g] Var(X)$$

$$eval[g'(x)]^2 @ E[x]$$

~~$$Var(g(\hat{x})) = \Delta_\theta^2 Var(\hat{\theta}) + \Delta_\alpha^2 Var(\hat{\alpha}) + 2 \Delta_\theta \Delta_\alpha Cov(\hat{\theta}, \hat{\alpha})$$~~

Binomial-Beta

$$N \sim Bin(m, q)$$

$$q \sim Beta(a, b)$$

read $F(x)$ CAREFULLY (denom x)

w/ x claims

n exposures
 den in m d

POSTERIOR

$$q \sim Beta$$

$$a^* = a + x$$

$$b^* = b + m - x$$

non-claims

Normal-Normal

$$X \sim Normal(\theta, v)$$

$$\theta \sim Normal(\mu, a)$$

$$v = E[Var(x)]$$

$$a = Var(E[x])$$

Also Follo
 IMPORTANT

For $\ln-N$, translate
 $\ln$ to N

$$P_{\hat{x}} \sim Normal(\mu, (1-z)a + v)$$

→ works for 0 loss
 in 0 years
 too

~~$$P_{\hat{x}} \sim Normal(\mu + z(\bar{x} - \mu), (1-z)a + v)$$~~

Empirical Bayes 3 Step Summary

① All non-par use ugly v, μ

② All semi-par use $\hat{v} = \hat{\mu} = \bar{x}$

③ Semi-par $\Pi(\mu)$ has simple $\hat{a} = s^2 - \hat{v}$

important if denom
 $a=0$

$$\hat{a} = s^2 - \hat{v}$$

$$\uparrow \quad \uparrow \quad \downarrow$$

$$\frac{1}{n+1} \quad \frac{1}{n} \quad \text{for } n \text{ par}$$

$$\chi^2 = \sum \frac{(O-E)^2}{E}$$

EXCEPT

if y_r is a column

$$\rightarrow \sum \frac{(O-E)^2}{V}$$

$$y_r / n \times$$

Minimize

$L(\theta) = MSE \rightarrow$ mean

$L(\theta) = |\hat{\theta} - \theta| \rightarrow$ median

Coherence - "Expect coherent tails"

$E[X]$ Yes

$TVAR$ Yes

• Splice

"Continuous" $\rightarrow f^-(x) = f(x)$

• $X \sim \exp(\theta)$

$$\sum_1^{\hat{n}} X \sim \text{Gamma}(\alpha = n, \theta)$$

MSE & bias
has FIRST

• $E[(\hat{\theta} - \theta)^2 | \theta]$ if $X \sim \text{Pareto}(\alpha = 5, \theta)$
 $\hat{\theta} = 6X^2$

$$= E[(6X - \theta)^2 | \theta]$$

$$= E[36X^2 - 12X^2\theta + \theta^2 | \theta]$$

$$= 36E[X^2] - 12E[X^2]\theta + \theta^2$$

$$g(x) = a^x$$

$$g'(x) = \ln a \cdot a^x$$

$$\text{Cov}(X_1, X_2) = a$$

• CI for $\text{Var}_p(x)$

$$a, b = nq \pm .5 \pm z \sqrt{nq(1-q)}$$

round down round up

• Do NOT apply v adj for del to N,
apply to PARAMETER: $N \sim \text{NB}(r, \text{VB})$

MLE = MoM for:

"Poisson a NEGATIVE ENGLISH individual"

Poisson
Neg Bin (No grouped data)
Exp
Normal
Gamma

• " χ^2 is more robust w/o adj"

Only χ^2 can handle groups, discrete, auto adj when fitting

• Inflation into scale parameter θ

• "Maximum Covered Loss" refers to ground-up
NOT max PMT

• For $\text{Var}((X-d)_+)$ of $X \sim \exp(\theta)$

$$\text{Set } (X-d)_+ = \begin{cases} 0 & X < d \\ X-d & X > d \end{cases}$$

$$\text{Var}[(X-d)_+] = E[\text{Var}(Y|\theta)] + \text{Var}(E[Y|\theta])$$

$$\stackrel{\text{exp}}{=} \theta^2 P(X > d) + (\theta - 0)^2 P(X < d) P(X > d)$$

$$\int e^{xu} \cdot f(x) dx = M(u)$$

↑
MGF @ u

• Normal approx

↳ continuity correction!
↳ to match ABCDE

• Log Likelihood Test

Reject H_0 for H_1

$$\text{if } 2(\ell_1 - \ell_0) > \chi^2_{df=\# \text{ new param}}$$

(note: ℓ_0 & ℓ_1 will be neg - there's diff!)
↳ log-likelihood

Go through models as "king of the hill" starting w/ LEAST param.

* Want high
(neg log-likelihood)
↓
high likelihood

"entries occur @ interval boundaries"

$$\begin{matrix} d_j @ \text{beg} \Rightarrow su & d_0 @ 0 \\ u_j @ \text{end} & u_0 @ 1 \end{matrix}$$

• Never do quadratics with $a < 0$ (get pos a)

• Profit problems: set up # years & little details CAREFULLY

• "Bühlshitt": Bühl approx Bayes

• The — Translated Invariance $P(x+c) = P(x) + c$
Properties — Positive Homogeneity $P(cx) = cP(x)$
For Measuring — Monotonicity For $x \leq y$, then $P(x) \leq P(y)$ ← SS fails
Shit — Subadditivity $P(x+y) \leq P(x) + P(y)$ ← Var fails

$$\hat{\theta}_{MLE} \stackrel{\text{exp}}{=} \frac{\$}{\# \text{ uncensored losses}}$$

(Truly \$, even with ded's)

Comparing Fit Tests

Like χ^2 all
"Smirnov-Darling" is needed

K-S	A-D	χ^2	Likelihood Ratio
Ind	Ind	Ind & Group	Ind & Group
Censor: Crit ↓	Censor: Crit ↓	No change	No change
$n \rightarrow \infty \Rightarrow$ Crit ↓	No change	No change	No change
Fixed: Crit ↓ date	Fixed: Crit ↓ date	No change	No change
	More weight on tails	More weight on low prob group	

Darling is good
↓
no irrelevance
& ger tail

• PGF $P'(0) = p_1$

$P(z) = E[z^x]$

$P'(1) = E[x]$

$P''(1) = E[x(x-1)]$

$P'''(1) = E[x(x-1)(x-2)]$

⋮

⋮

Bootstrap
• $MSE = E_{\text{emp}} \left[\left(\overset{\text{varies}}{\theta_i} - \overset{\text{constant}}{\theta_{\text{emp}}} \right)^2 \right]$

• Stupid estimators ONLY for θ_i ,
 θ_{emp} is really the median/real value

• $TVAR = E[L | L > \pi_p]$
 $= E[L - \pi_p | L > \pi_p] + \pi_p$

$TVAR = e_L(\pi_p) + \pi_p$
 $= \frac{E[L] - E[L \wedge \pi_p]}{1-p} + \pi_p$

• $TVAR(aX-b) = aTVAR(X) - b$

Bühl-Bayes

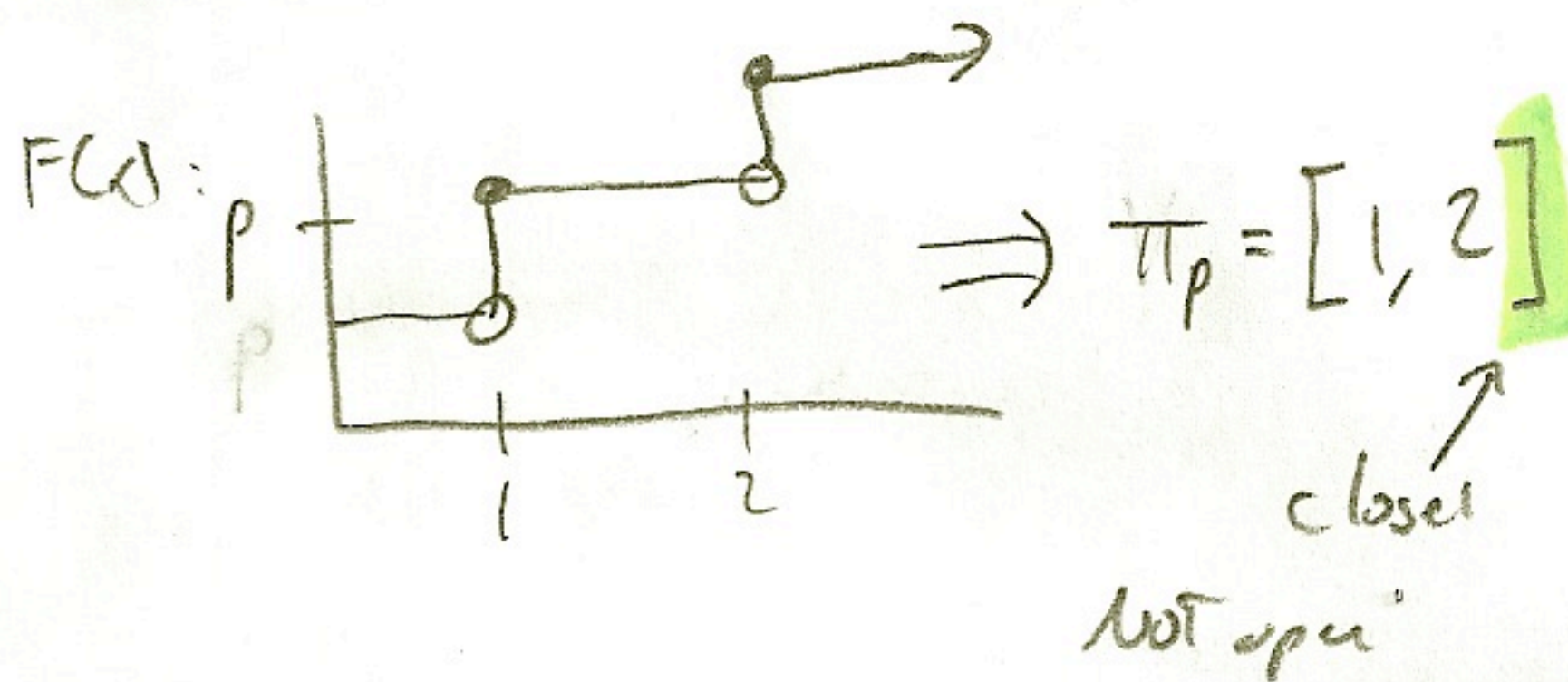
For any conj prior situations

Claims experience & uni X

Bounds change to make sense!

Percentile

Anywhere $F(x)$ equals OR jumps over



Bootstrap (special case)

MSE of Mean: $\frac{\sum (x - \bar{x})^2}{n^2}$

$MSE = E \left[\left(\underset{\text{varies}}{\hat{\theta}_i} - \underset{\text{constant}}{\theta} \right)^2 \right]$
plugin

① Translational Invariance

$P(x+c) = P(x) + c$

② Positive Homogeneity

$P(cx) = c \cdot P(x)$

③ Monotonicity

$x < y \rightarrow P(x) < P(y)$

④ Subadditivity

$P(x+y) \leq P(x) + P(y)$

"Together is not more"
↓
No synergy