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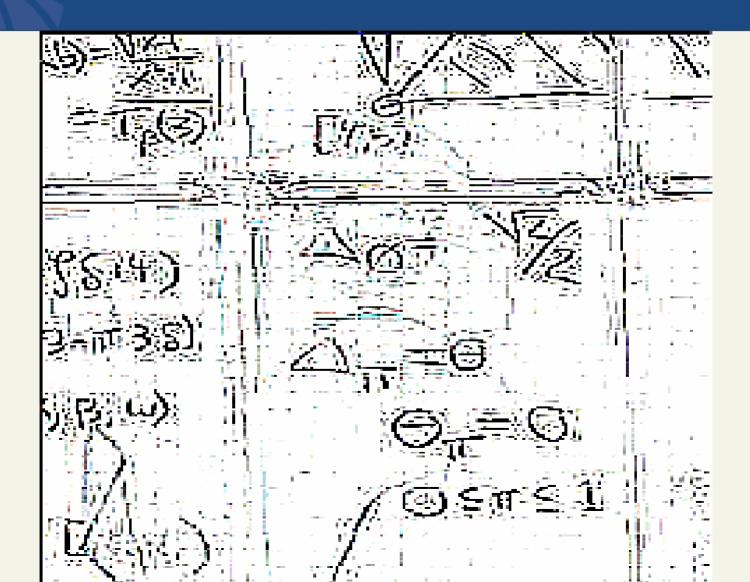
THE WEIGHTING GAME

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Getting rid of the superfluous information



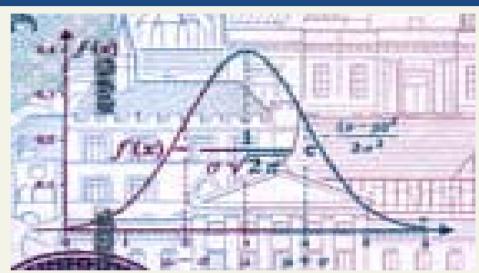


How the presentation could have started, but didn't: Proof that statisticians can speak alien languages

• Let (Ω, K, P) be a probability space, where (Ω, K) is a measurable space and $P: K \mapsto [0,1]$

is a probability measure function from the σ -algebra K. It is perfectly natural to ask oneself what a σ -algebra or σ -field is.

Definition. A σ -field is a collection of subsets K of the sample space Ω with



$$\Phi \in K$$
, $A \in K \Rightarrow A^{C} \in K$, $A_{1}, A_{2}, ... \in K \Rightarrow \bigcup A_{i} \in K$

Of course, once we mastered the σ -algebra or σ -field concept it is only reasonable to wonder what a probability measure is

Definition. A probability measure P has the following properties

Where do all these fit in the big picture?

Every sample space is a particular case of probability space and weighting is intrinsically related to sampling



$$P(\Omega) = 1, A_1, A_2, ... \in K \text{ disjoint sets} \Rightarrow P(\bigcup A_i) = \sum P(A_i)$$

Why simple questions can have complex answers?

Question: What is the average length of in-hospital stay for patients?

Complexity: The original question is imprecise.

New question: What is the average length of stay for:

- Several hospitals of interest?
- Maryland hospitals?
- Blue State hospitals? ...



"Data" Collection & Goal

Survey, conducted in 5 hospitals

- Hospitals are selected
- n_{hospital} patients are sampled at random
- Length of stay (LOS) is recorded
- Goal: Estimate the population mean



Procedure

- Compute hospital specific means
- "Average" them
 - For simplicity assume that the population variance is known and the same for all hospitals
- How should we compute the average?
- Need a (good, best?) way to combine information



DATA

Hospital	# sampled n _{hosp}	Hospital size	% of Total size: $100\pi_{\text{hosp}}$	Mean LOS	Sampling variance
1	30	100	10	25	$\sigma^{2}/30$
2	60	150	15	35	$\sigma^2/60$
3	15	200	20	15	$\sigma^{2}/15$
4	30	250	25	40	$\sigma^2/30$
5	15	300	30	10	$\sigma^{2}/15$
Total	150	1000	100		



Weighted averages

Examples of various weighted averages: $\sum w_i \overline{X}_i$

Weighting strategy	Weights x100	Mean	Variance Ratio
Equal	20 20 20 20 20	25.0	130
Inverse variance	20 40 10 20 10	29.5	100
Population	10 15 20 25 30	23.8	1 <mark>72</mark>

Variance using inverse variance weights is smallest



What is weighting? (via Constantine)

- Εσσενχε: α γενεραλ ωαψ οφ χομπυτινγ απεραγεσ
- Τηερε αρε μυλτιπλε ωειγητινγ σχηεμεσ
- Μινιμιζε παριανχε βψ υσινγ ινπερσε παριανχε ωειγητσ
- Μινιμιζε βιασ φορ τηε ποπυλατιον μεαν
- Πολιχψ ωειγητσ



What is weighting?

- The Essence: a general (fancier?) way of computing averages
- There are multiple weighting schemes
- Minimize variance by using inverse variance weights
- Minimize bias for the population mean by using population weights ("survey weights")
- Policy weights
- "My weights," ...



Weights and their properties

- Let $(\mu_1, \mu_2, \mu_3, \mu_4, \mu_5)$ be the TRUE hospital-specific LOS
- Then $\sum_{w_i} \overline{X}_i$ estimates $\sum_{w_i} \mu_i$
- If $\mu_1 = \mu_2 = \mu_3 = \mu_4 = \mu_5 = \mu_\pi = \sum \mu_i \pi_i$ ANY set of weights that add to 1 estimate μ_π .
- So, it's best to minimize the variance
- But, if the TRUE hospital-specific E(LOS) are not equal
 - Each set of weights estimates a different target
 - Minimizing variance might not be "best"
 - An unbiased estimate of μ_{π} sets $\mathbf{w}_{i} = \pi_{i}$

General idea

Trade-off variance inflation & bias reduction



Mean Squared Error

General idea

Trade-off variance inflation & bias reduction

MSE = Expected(Estimate - True)² = Variance + Bias²

- Bias is unknown unless we know the μ_i (the true hospital-specific mean LOS)
- But, we can study MSE (μ , w, π)
- Consider a true value of the variance of the between hospital means
- Study BIAS, Variance, MSE for various assumed values of this variance



Mean Squared Error

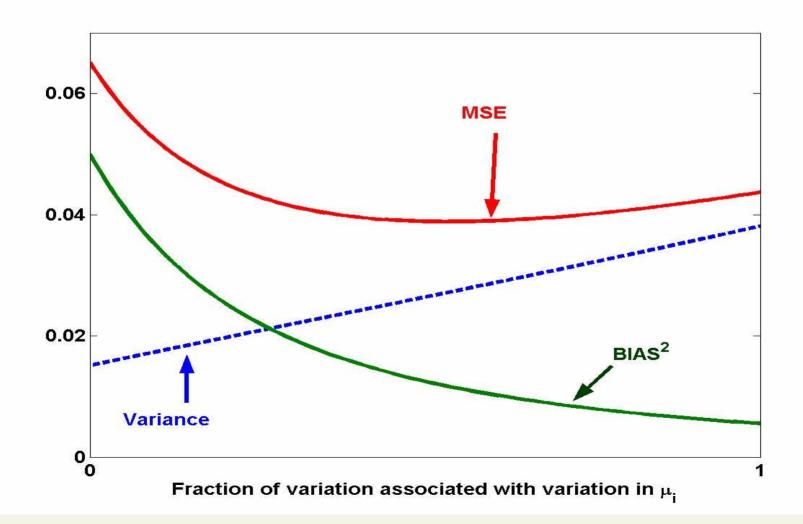
 Consider a true value of the variance of the between hospital means

$$T = \Sigma(\mu_i - \mu^*)^2$$

- Study BIAS, Variance, MSE for optimal weights based on assumed values (A) of this variance
- When A = T, MSE is minimized
- Convert T and A to fraction of total variance



The bias-variance trade-off X is assumed variance fraction Y is performance computed under the true fraction





Summary

- Much of statistics depends on weighted averages
- Choice of weights should depend on assumptions and goals
- If you trust your (regression) model,
 - Then, minimize the variance, using "optimal" weights
 - This generalizes the equal μ s case
- If you worry about model validity (bias for μ_{π}),
 - Buy full insurance, by using population weights
 - You pay in variance (efficiency)
 - Consider purchasing only what you need

Using compromise weights



Statistics is/are everywhere!





EURO: our short wish list



