Statistical Modeling: Bigger and Bigger

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"in data analysis there is no loner any problem of computation"

- Benzécri, 2005

# **Logistic Regression**

•Linear model for log odds of category membership:

$$\log \frac{p(y=1 | \mathbf{x}_i)}{p(y=-1 | \mathbf{x}_i)} = \sum \beta_j x_{ij} = \beta \mathbf{x}_i$$

# Maximum Likelihood Training

 Choose parameters (β<sub>j</sub>'s) that maximize probability (likelihood) of class labels (y<sub>i</sub>'s) given documents (x<sub>i</sub>'s)

$$L(oldsymbol{eta}) = p(oldsymbol{eta}|D) = (\prod_{i=1}^n rac{1}{1+\exp(-oldsymbol{eta}^Toldsymbol{x}_iy_i)})$$

- Tends to overfit
- Not defined if *d* > *n*
- Feature selection

## Shrinkage Methods

- Avoid combinatorial challenge of feature selection
- L1 shrinkage: regularization + feature selection
- Expanding theoretical understanding
- Large scale
- Empirical performance

**Ridge Logistic Regression** 

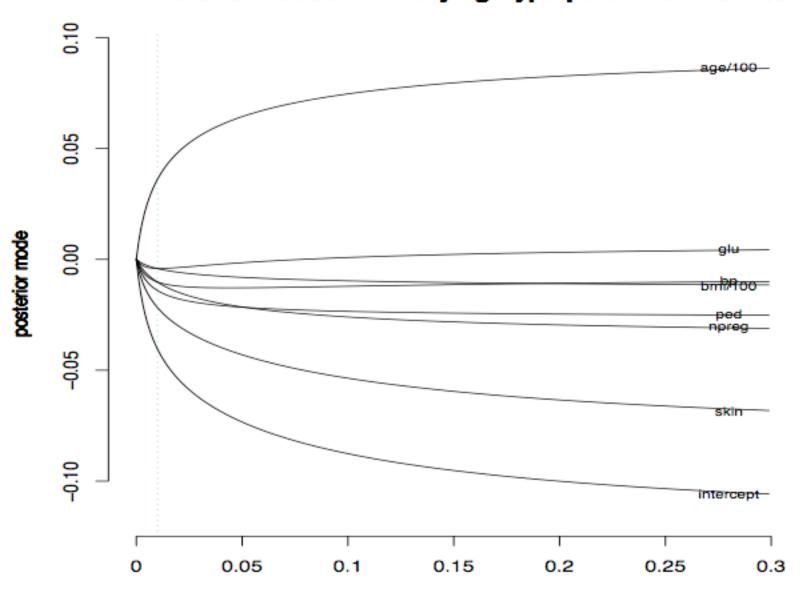
Maximum likelihood plus a constraint:

$$\sum_{j=1}^p \beta_j^2 \le s$$

#### Lasso Logistic Regression

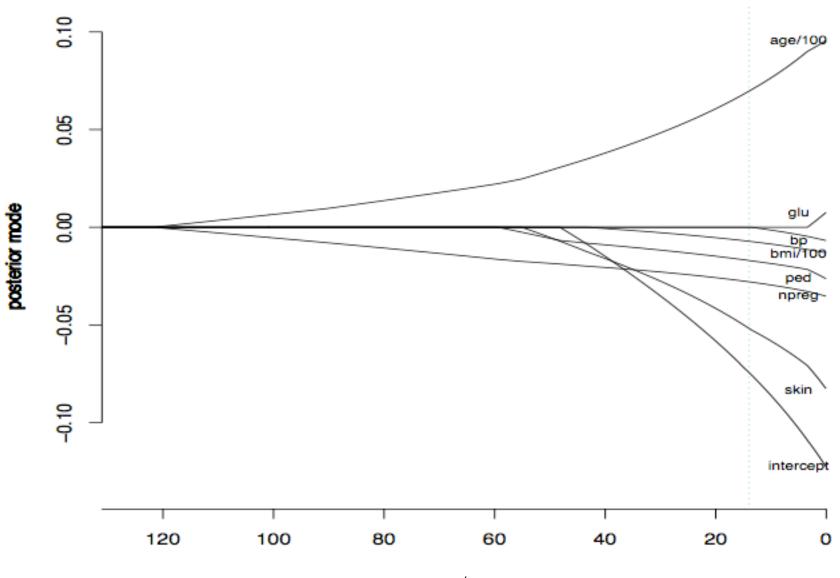
Maximum likelihood plus a constraint:

$$\sum_{j=1}^{p} \left| \beta_{j} \right| \le s$$



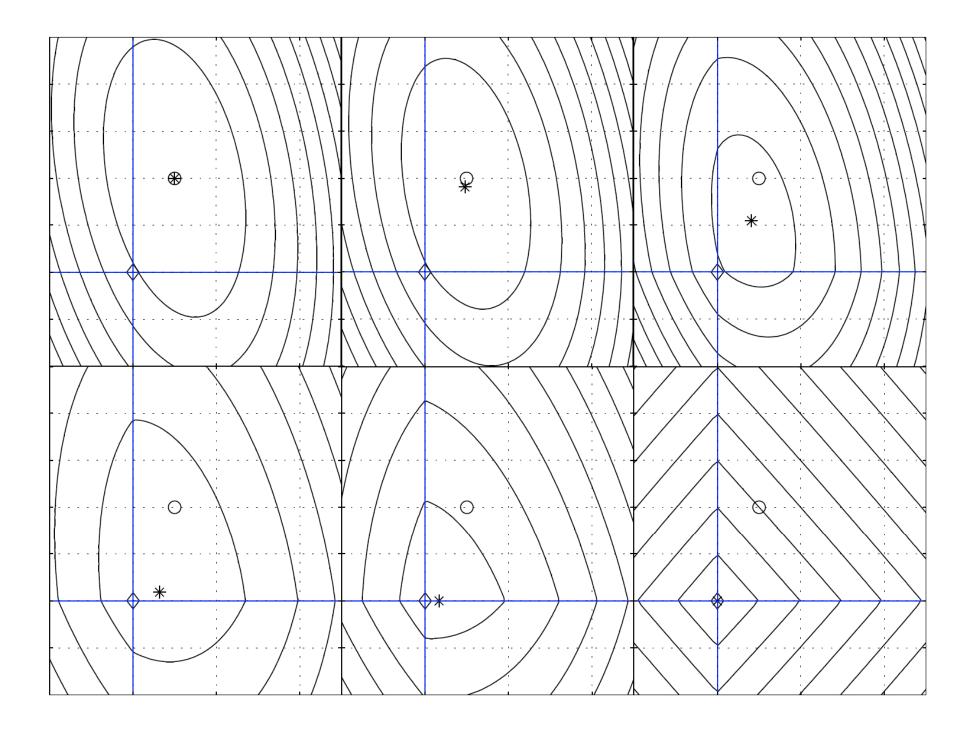
Posterior Modes with Varying Hyperparameter – Gaussian

S

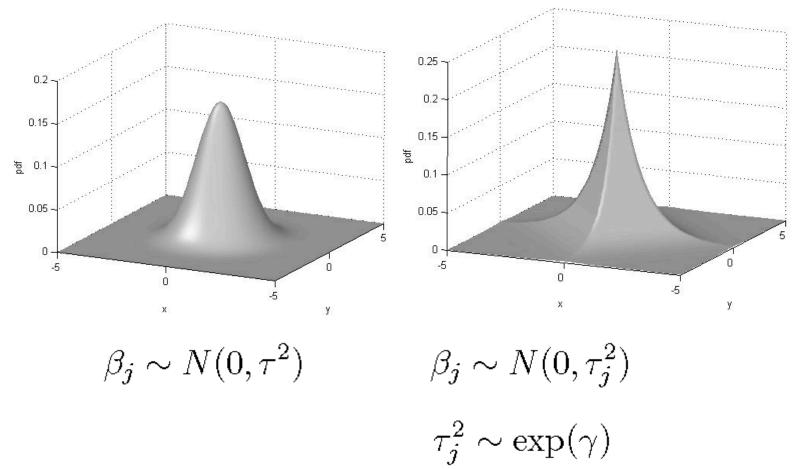


Posterior Modes with Varying Hyperparameter – Laplace

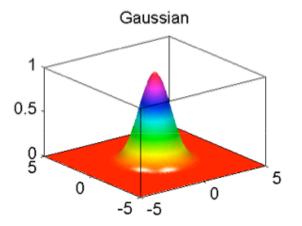
1/s

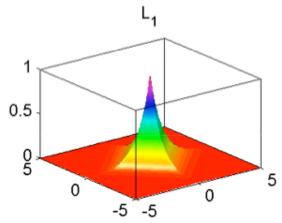


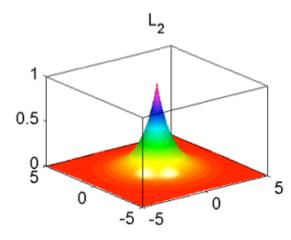
## Bayesian Perspective

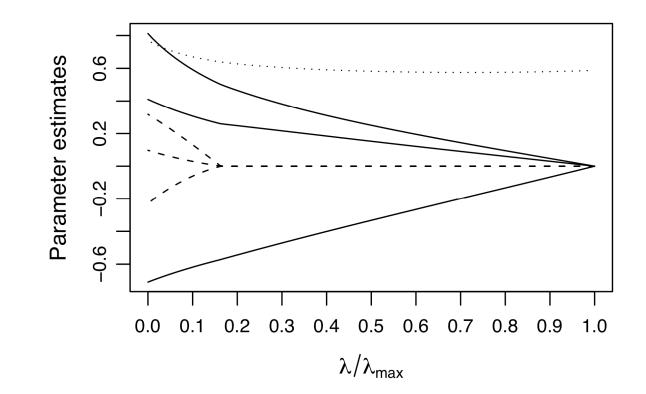




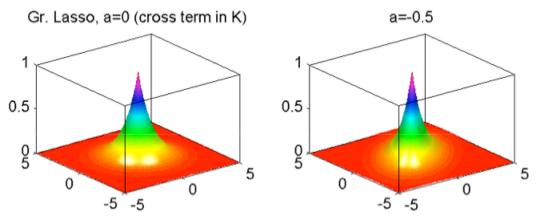




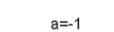


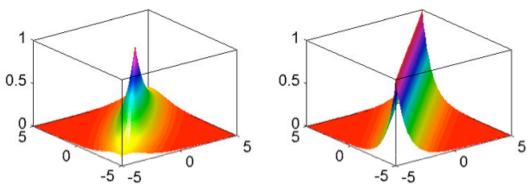


#### "soft fusion"









Balakrishnan and Madigan (2007)

## "Consistency"

- Lasso not always consistent for variable selection
- SCAD (Fan and Li, 2001, JASA) consistent but nonconvex
- relaxed lasso (Meinshausen and Buhlmann), adaptive lasso (Wang et al) have certain consistency results
- Zhao and Yu (2006) "irrepresentable condition"

#### Implementation

- Open source C++ implementation. Compiled versions for Linux, Windows, and Mac (soon)
- Binary and multiclass, hierarchical, informative priors
- Gauss-Seidel co-ordinate descent algorithm
- Fast? (parallel?)
- http://stat.rutgers.edu/~madigan/BBR

## Aleks Jakulin's results

domain	BMR			s / ins TAN		RKT	BK3
krkp	0.09			0.19			0.05
monk2	0.65			0.63			0.45
tic-tac-toe	0.03			0.49			0.07
titanic		-0.53			0.48		
lenses		0.72				and the second s	
monk1		0.49			0.01		0.02
mushroom	0.00			0.00	0.01		0.02
shuttle		0.10			0.07		
soy-small*		.0.31			0.00		0.07 0.00
		0.09					
wine yeast-class*		0.05		_	0.19		0.11
-				0.03			0.12
anneal		0.05					0.11
balance-scale		0.17					0.51
lung-cancer*		1.02					1.18
monk3		0.11				0.11	
post-op		0.61		1.78			0.67
promoters*		0.23		-			0.52
adult							0.30
audiology*		1.31		$\cdot 5.56$			
australian	0.33			·0.94		0.37	
breast-LJ	0.55			0.89			0.58
breast-wisc	0.10	0.12			0.21		0.16
bupa	0.60	0.60		0.60	0.62		0.61
car	0.18			0.18	0.19		0.19
cinc	0.91			-1.03	- Martin Control State		0.92
crx	0.33			0.93			
ecoli	0.45			·0.94			0.81
german	0.50	0.51		·1.04		and the second s	0.59
glass	0.74			·1.76	1.12		0.99
hayes-roth	0.29			-1.18	0.45	0.45	0.45
heart	1.01	1.03	1.25	$\cdot 1.53$	1.11	1.09	1.09
hepatitis	0.36			$\cdot 1.31$		0.39	0.39
horse-colic	0.71	0.71	1.67	-5.97	0.83	0.82	0.82
ionosphere	0.19	0.26	0.64	$\cdot 0.74$	0.39	0.30	0.30
iris	0.16	0.24	0.27	0.32	0.27	0.18	0.18
lymph	0.50	0.56	1.10	$\cdot 1.25$	0.98	0.79	0.79
o-ring	0.66	0.80	0.83	0.76	1.41	0.67	0.67
p-tumor*	1.82	1.93	3.17	-4.76	2.65	2.55	2.55
pima	0.46	0.48	0.50	<u>0.49</u>	0.51	0.48	0.48
segment	0.13			-1.06			0.17
soy-large*	0.25	0.46	0.57	0.47	0.68	0.66	0.66
spam	0.15		0.53		0.19		0.19
vehicle	0.54		1.78		0.69	0.66	0.66
voting	0.11		0.60				0.14
wdbc	0.09			0.29			0.13
200*	0.35				0.40		0.38
avg rank	2.13		5.62	5.60	4.74	3.68	3.36

## 1-of-K Sample Results: brittany-l

Feature Set	% errors	Number o Features	of	
"Argamon" function words, raw tf	74.8	380		
POS	75.1	44		
1suff	64.2	121		
1suff*POS	50.9	554		
2suff	40.6	1849		
2suff*POS	34.9	3655	<b>4.6</b> I	million parameters
3suff	28.7	8676	X	/
3suff*POS	27.9	12976		
3suff+POS+3suff*POS+Arga mon	27.6	22057		
All words	23.9	52492 📕		

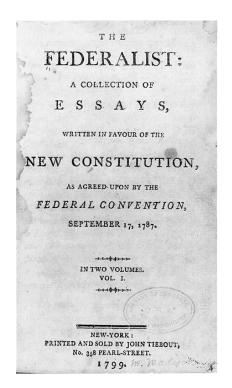
89 authors with at least 50 postings. 10,076 training documents, 3,322 test documents.

BMR-Laplace classification, default hyperparameter

Madigan et al. (2005)

## The Federalist

- "The authorship of certain numbers of the 'Federalist' • has fairly reached the dignity of a well-established historical controversy." (Henry Cabot Lodge, 1886)
- Historical evidence is muddled •



Paper Number	Author
1	Hamilton
2-5	Jay
6-9	Hamilton
10	Madison
11-13	Hamilton
14	Madison
15-17	Hamilton
18-20	Joint: Hamilton and Madison
21-36	Hamilton
37-48	Madison
49-58	Disputed
59-61	Hamilton
62-63	Disputed
64	Jay
65-85	Hamilton



#### JOURNAL OF THE AMERICAN STATISTICAL ASSOCIATION

Number 302

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Volume 58

#### INFERENCE IN AN AUTHORSHIP PROBLEM<sup>1,2</sup>

A comparative study of discrimination methods applied to the authorship of the disputed *Federalist* papers

FREDERICK MOSTELLER Harvard University and Center for Advanced Study in the Behavioral Sciences AND DAVID L. WALLACE University of Chicago

•Used function words with Naïve Bayes with Poisson and Negative Binomial model

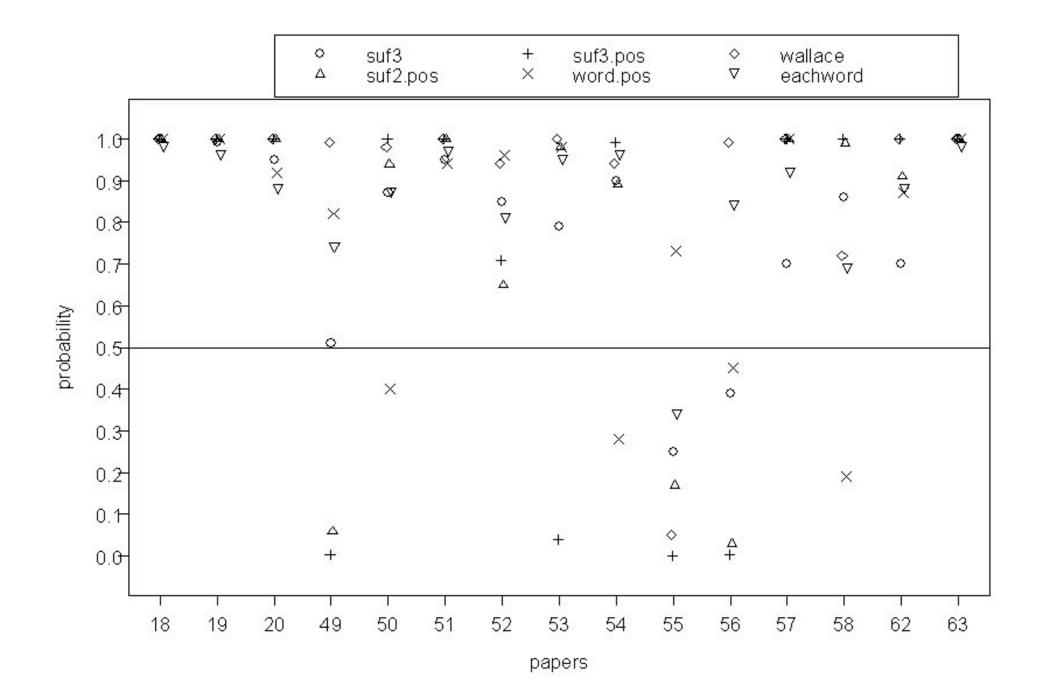
Out-of-sample predictive performance

#### F. Summing up

In summary, the following points are clear:

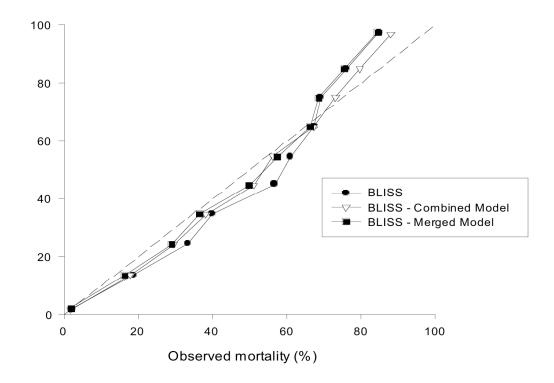
1) Madison is the principal author. These data make it possible to say far more than ever before that the odds are enormously high that Madison wrote the 12 disputed papers. Weakest support is given for No. 55. Support for Nos. 62 and 63, most in doubt by current historians, is tremendous.

Feature Set	10-fold Error Rate	
Charcount	0.21	
POS	0.19	
Suffix2	0.12	
Suffix3	0.09	
Words	0.10	
Charcount+POS	0.12	
Suffix2+POS	0.08	
Suffix3+POS	0.04	best
Words+POS	0.08	
484 features	0.05	
Wallace features	0.05	
Words (>=2)	0.05	
Each Word	0.05	



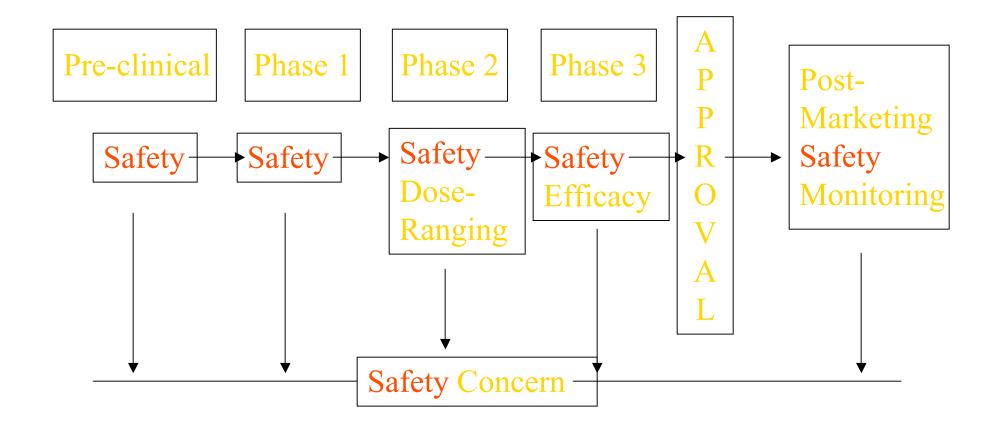
#### **Risk Severity Score for Trauma**

- Standard "ICISS" score poorly calibrated
- Lasso logistic regression with 2.5M predictors:



Burd and Madigan (2006)

#### Safety in Lifecycle of a Drug/Biologic product



## Databases of Spontaneous ADRs

- FDA Adverse Event Reporting System (AERS)
  - Online 1997 replace the SRS
  - Over 250,000 ADRs reports annually
  - 15,000 drugs 16,000 ADRs
- CDC/FDA Vaccine Adverse Events (VAERS)
  - Initiated in 1990
  - 12,000 reports per year
  - 50 vaccines and 700 adverse events
- Other SRS
  - WHO international pharmacovigilance program

U.S. Department of Health and Human Services		Form Approved:	OMB No. 0910-0291, Expires: 10/31/08
MEDWATCH	For VOLUNTARY reporting of		See OMB statement on reverse. DA USE ONLY
MedWatch	adverse events, product problems	and Triage unit	DA USE ONLY
The FDA Safety Information and	product use errors	sequence #	
Adverse Event Reporting Program	Page of		
A. PATIENT INFORMATION	D. SUSPECT	PRODUCT(S)	
1. Patient Identifier 2. Age at Time of Event, or 3. Sex		h, Manufacturer (from product label)	1
Date of Birth: Fer		,,	
In confidence			
B. ADVERSE EVENT, PRODUCT PROBLEM OR	EBBOR #2		
Check all that apply:	2. Dose or Am	ount Frequency	Route
1. Adverse Event Product Problem (e.g., defectation	#1		
Product Use Error Problem with Different Manufacto			
2. Outcomes Attributed to Adverse Event	#2		
(Check all that apply)	3. Dates of Use (	If unknown, give duration) from/to (or	5. Event Abated After Use
Death: Disability or Pe	manent Damage best estimate)		Stopped or Dose Reduced?
(mm/dd/)yyyy) Life-threatening Congenital And	maly/Birth Defect		A1 Yes No Doesn't Apply
	Important Medical Events) 42		A2 Yes No Doesn't
	d Disesseis of F	leason for Use (Indication)	Арру
Required Intervention to Prevent Permanent Impairment/Dat			<ol> <li>Event Reappeared After Reintroduction?</li> </ol>
3. Date of Event (mm/dd)(yyy) 4. Date of this Rep	ort (mm/da/yyyy)		At Vec No Doesn't
	#2		Apply
5. Describe Event, Problem or Product Use Error	6. Lot#	7. Expiration Date	42 Yes No Doesn't Apply
	#1	#1	9. NDC # or Unique ID
	#2	A2	
		MEDICAL DEVICE	
<u></u>	1. Brand Name	MEDICAL DEVICE	
3			
	2. Common Devi	be Name	
	2. Manufactures	Name, City and State	
	3. Wanunacturer	tame, city and State	
	4. Model #	Lot #	5. Operator of Device
5			Health Professional
4	Catalog #	Expiration Date (	nmidd)yyy)
	Serial #	Other #	Other:
	6. If implemented	for Date (considering) 7 H E	minuted. One Pate immittinged
	6. If Implanted, G	ive Date (mm/dd/yyyy) 7. If Ex	cplanted, Give Date (mm/dd)(yyy)
	8. Is this a Single	use Device that was Reprocessed	and Reused on a Patient?
		No	
	9. If Yes to item ?	vo. 8, Enter Name and Address of I	Reprocessor
6. Relevant Tests/Laboratory Data, Including Dates			
	F. OTHER (	ONCOMITANT) MEDICAL	PRODUCTS
		and therapy dates (exclude treatment	
<ol> <li>Other Relevant History, Including Preexisting Medical Condi race, pregnancy, smoking and alcohol use, liver/kidney problems</li> </ol>	dons (e.g., allergies, ; etc.) C DEPODT	R (See confidentiality see	etion on back)
	1. Name and Add		don on backy
	Phone #	E-mail	
C. PRODUCT AVAILABILITY	2. Health Profess	ional? 3. Occupation	4. Also Reported to:
Product Available for Evaluation? (Do not send product to FDA)	Yes 🗌	No	Manufacturer
	5. If you do NOT	want your identity disclosed	User Facility
Yes No Returned to Manufacturer or:	(mm/dd/yyyy) to the manufact	turer, place an "X" in this box:	Distributor/Importer

FORM FDA 3500 (10/05) Submission of a report does not constitute an admission that medical personnel or the product caused or contributed to the event.

## Weakness of SRS Data

- Passive surveillance
  - Underreporting
- Lack of accurate "denominator", only "numerator"
  - "Numerator": No. of reports of suspected reaction
  - "Denominator": No. of doses of administered drug
- No certainty that a reported reaction was causal
- Missing, inaccurate or duplicated data

# Existing Methods

- Multi-item Gamma Poisson Shrinker (MGPS)
  - US Food and Drug Administration (FDA)
- Bayesian Confidence Propagation Neural Network
  - WHO Uppsala Monitoring Centre (UMC)
- Proportional Reporting Ratio (PRR and aPRR)
  - UK Medicines Control Agency (MCA)
- Reporting Odds Ratios and Incidence Rate Ratios
  - Other national spontaneous reporting centers and drug safety research units

# Existing Methods (Cont'd)

• Focus on 2X2 contingency table projections

	AE j = Yes	AE j = No	Total
Drug <i>i</i> = Yes	<i>a</i> =20	<i>b</i> =100	120
Drug <i>i</i> = Yes Drug <i>i</i> = No Total	<i>c</i> =100	<i>d</i> =980	1080
Total	120	1080	1200

- 15,000 drugs \* 16,000 AEs = 240 million tables
- Most  $N_{ij} = 0$ , even though N.. very large

## The Different Measures

Measure of Association	Formula	Probabilistic Interpretation
RR Relative Risk*	$\frac{a}{(a+b+c+d)}$ (a+c)*(a+b)	$\frac{\Pr(ae \mid drug)}{\Pr(ae)}$
PRR Proportional Reporting Ratio	a / (a + b) c / (c + d)	$\frac{\Pr(ae \mid drug)}{\Pr(ae \mid \neg drug)}$
ROR Reporting Odds Ratio	a / c  b / d	$\frac{\Pr(ae \mid drug) / \Pr(\neg ae \mid drug)}{\Pr(ae \mid \neg drug) / \Pr(\neg ae \mid drug)}$
Information Component	Log 2 $\frac{a}{(a + b + c + d)}$ $\frac{a}{(a + c)} (a + b)$	$\log_2 \frac{\Pr(ae \mid drug)}{\Pr(ae)}$

# Relative Reporting Ratio $(RR_{ij}=N_{ij}/E_{ij})$ • Advantages- Simple- Easy to interpret• Disadvantages• Disadvantages

 $N_{\dot{r}}$ 

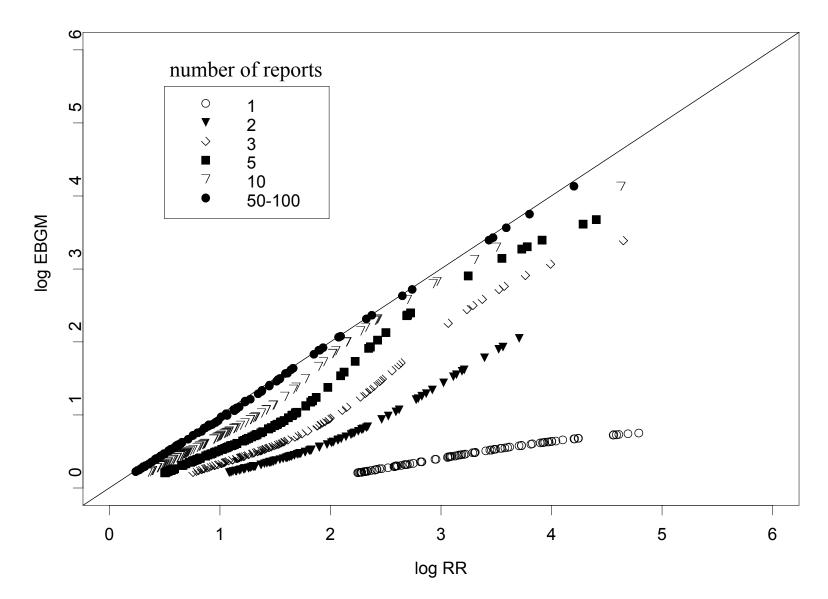
N..

 Extreme sampling variability when baseline and observed frequencies are small

(N=1, E=0.01 v.s. N=100, E=1)

- GPS provides a shrinkage estimate of RR that addresses this concern.

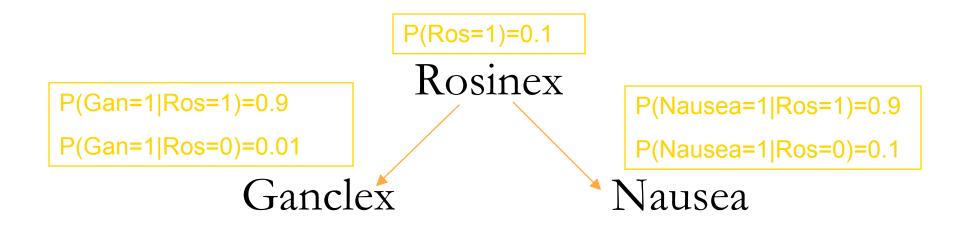
#### GPS SHRINKAGE – AERS DATA



# Confounding

• Contingency table analysis ignores effects of drugdrug association on drug-AE association

	Ros	inex	No R	osinex	Тс	otal	Ganclex
	Nausea	No Nausea	Nausea	No Nausea	Nausea	No Nausea	
Ganclex	81	9	1	9	82	18	Rosinex
No Ganclex	9	1	90	810	99	811	
RR		1		1	4.	58	Nausea



		Nausea vs. Ganclex		Nausea vs. Rosinex	
		Value	Rank	Value	Rank
N	[	1673	2	1826 1	
Bayesian Logistic Regression	Laplace-CV	0.0	9127	4.0	7
GPS E	BGM	2.8	73	3.0	68
Observ	ed RR	2.8	744	3.0	681

# Logistic Regression

- log [P/(1-P)] = intercept + ∑ (each drug effect)
   P = Pr (report with these drugs will have the AE)
- 15,000 logistic regressions with  $n \approx 3$  million
- 15,000 main effects
- millions of pairwise interactions???

## Current Work

- Model associations between *groups* of drugs and *groups* of adverse events
- Bayesian generative approach applicable
- Sketch:
  - assign every drug to a latent group
  - assign every AE to a latent group
  - for each set of drugs and set of AE's, generate a report with probability defined by latent group memberships
- Major computational challenges
- Blei, Feinberg, Ghahramani, Roweis, etc.



- BR has 5 APs, site dimension: 225 ft X 144 ft
  - 259 blue (corridor) data points taken earlier

## Hierarchical Model $Y_i$ $X_i$ $D_{ij}$ $S_{ij}$ *i*=1,...,*n* b<sub>1j</sub> b<sub>0j</sub> *j*=1,...,5 $au_0$ $\mu_0$ $au_1$ $\mu_1$