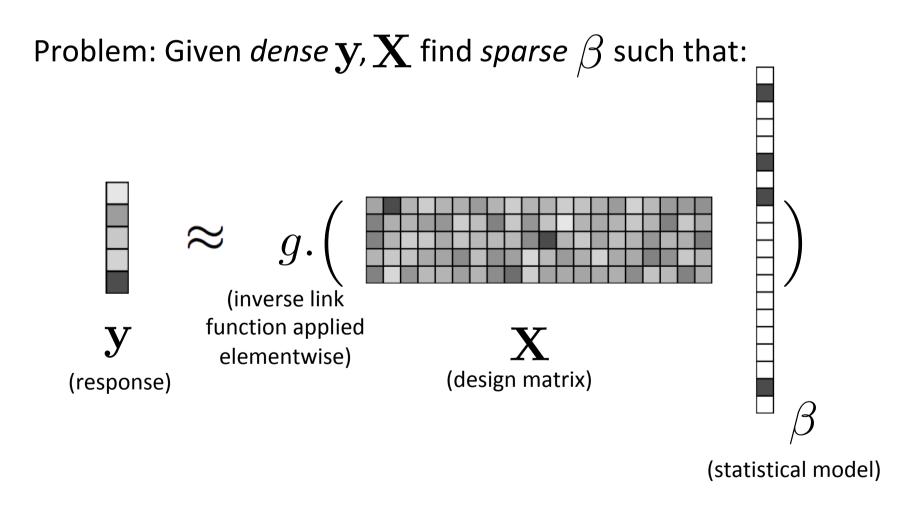
# **Project 1**: Iterative Hard Thresholding in GWAS: Generalized Linear Models, Prior Weights, and Double Sparsity



## IHT vs Lasso vs Marginal testing

	Normal	Logistic	Poisson	Neg Binomial
IHT TF	8.84	6.28	7.2	9.0
IHT FP	0.02	0.1	1.28	0.98
Lasso TP	9.52	8.16	9.28	NA
Lasso FP	31.26	45.76	102.24	NA
Marginal TP	7.18	5.76	9.04 (5.94*)	5.98
Marginal FP	0.06	0.02	1527.9 (0.0*)	0.0

TP = true positives (higher means good)

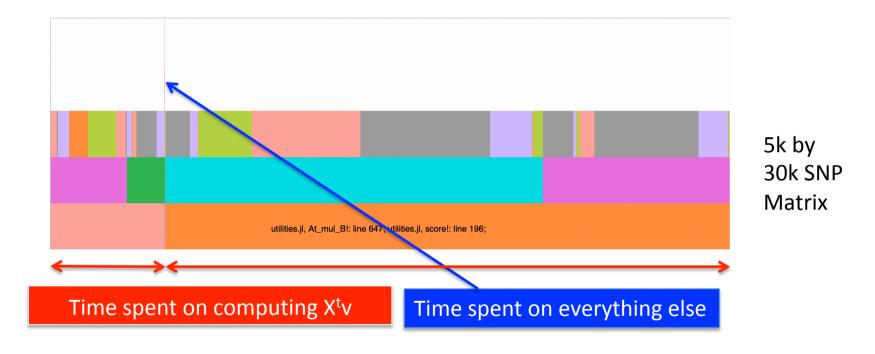
FP = false positives (lower means good)

\* = zero-inflated Poisson regression

**Other methods for high dimensional GLM**: Elastic net, MCP, Matching pursuit, stability selection, forward stepwise regression...

### Bottleneck of IHT is gradient computation

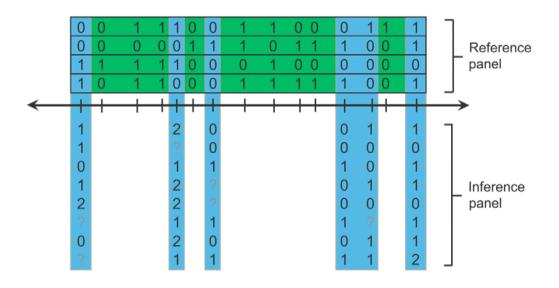
IHT iterates according to: 
$$\beta_{n+1} = P_{S_k} (\beta_n - s_n \nabla f(\beta_n))$$



Question: Can mailman algorithm (which appears to require a lot of preprocessing) help us?

#### **Project 2**: Genotype imputation via haplotype reference panels

Problem: Fill missing entries of the inference panel



T = SNPs typed in both panels

U = SNPs typed only in reference panel

Figure: Howie, Bryan N., Peter Donnelly, and Jonathan Marchini. "A flexible and accurate genotype imputation method for the next generation of genome-wide association studies." PLoS genetics 5.6 (2009): e1000529.

### Our method is based on least squares

Our competitors (Minimac4, Impute5, Beagle 5) all use Hidden Markov Models derived in (\*)

Preliminary results:		Java	C++	Julia
	Beagle 5.0 (phased)	Beagle 5.0 (unphased)	Minimac4 (phased)	Mendellmpute (unphased)
Error rate (%)	2.9	4.7	1.8	0.3
CPU time (sec)	11 (+2525)	1621	297 (+2525)	5.4

- 36,498 SNPs, 660 reference panels, 665 samples as imputation target
- Phasing data (beagle 4.1) adds 2525 seconds
- MendelImpute does not require pre-phasing
- Timing for MendelImpute does not include data import/export