Weak Signal Identification & Inference in Penalized Model Selection (Shi & Qu 2017) 1) Achieve better trade off bothern folise discovery X weak signal detection power @ finite sample than should be used instead of asymptotically. for weak ---Basics about Adaptive LASSO: $\hat{\theta}_{ALASSO} = (|\hat{\theta}^{LS}| - \frac{\lambda}{|\hat{\theta}^{LS}|})_{+} \operatorname{sgn}(\hat{\theta}^{LS})$ An > A if VIN In > 0, N In > 10. (order of In here in the ~ 1/2) noise Strength: C6 Why we should not. What signed asymptotically? Probability of Selection (for single variable): $Pa(\theta) = P(\theta_{ALASSo} \neq 0 \mid \theta_{Enum}) = \underline{\Phi}(\frac{\theta - \sqrt{\lambda}}{6/\sqrt{m}}) + \underline{\Phi}(\frac{-\theta - \sqrt{\lambda}}{6/\sqrt{m}})$.

We pair $(Y, y^{\gamma}): Y = \underline{\Phi}(\frac{y^{\gamma} - \sqrt{\lambda}}{6/\sqrt{m}}) + \underline{\Phi}(\frac{-\theta - \sqrt{\lambda}}{6/\sqrt{m}})$ signal should be at # least as strong as y^{γ} to be detected with p=Y. define transition phase:) $\theta \in \Theta^{(s)}$ $P_d > \gamma^s$ $\theta \in \Theta^{(u)}$ $\gamma^u < P_d < \gamma^s$ $P_d \leq \gamma^u$. good: 2) 0 bad 2 ~ Tan ~ (1 n = 1 , n = 4) slower than noise strongth. Detection: select 2, = 8 5 < T FDR tolerance. T> To, more detection. in finite sample case, U' is still too large to be ignored. $\mathcal{V}_{3} = \sqrt{1} + 2q \frac{6}{\sqrt{n}} \rightarrow lower bound for <math>Pd = 1 - \alpha$. (Why not apply to weak signals: no inference for undetected weak signals).

Inference: detected strong signals: asymptotic theory based interval by Zou. underestimated set se (to conservative for detected weak signal) -- .. weak -- : CI based on LSE (for full model).

Similarity with PoSI; only construct CI for selected porom.