

#### University of Luxembourg

THESIS FOR THE BACHELOR OF MATHEMATICS

# High Dimensional Regression Models

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## Contents

1	Notes			
	1.1	Section 6.2		
	1.2	Section 6.3	4	
_	Introduction			
	2.1	The first section	7	
		2.1.1 The first subsection	-	

iv CONTENTS

#### Chapter 1

#### Notes

#### 1.1 Section 6.2

$$\mathbf{Y} = \mathbf{X}\boldsymbol{\beta} + \boldsymbol{\varepsilon} \tag{1.1}$$

We define  $\hat{\beta}$  as follows

$$\hat{\beta} := \arg\min_{\beta} \left\{ \frac{\|\mathbf{Y} - \mathbf{X}\beta\|_2^2}{n} + \lambda \|\beta\|_1 \right\}$$
 (1.2)

Lemma 1.1 (Basic Inequality).

$$\frac{\|\mathbf{X}(\hat{\beta} - \beta^0)\|_2^2}{n} + \lambda \|\hat{\beta}\|_1 \le 2\frac{\varepsilon^T \mathbf{X}(\hat{\beta} - \beta^0)}{n} + \lambda \|\beta^0\|_1$$

*Proof.* By definition of  $\hat{\beta}$ , we have that

$$\forall \beta \quad \frac{\|\mathbf{Y} - \mathbf{X}\hat{\beta}\|_2^2}{n} + \lambda \|\hat{\beta}\|_1 \le \frac{\|\mathbf{Y} - \mathbf{X}\beta\|_2^2}{n} + \lambda \|\beta\|_1$$

In particular for  $\beta = \beta^0$  we have

$$\frac{\|\mathbf{Y} - \mathbf{X}\hat{\beta}\|_{2}^{2}}{n} + \lambda \|\hat{\beta}\|_{1} \leq \frac{\|\mathbf{Y} - \mathbf{X}\beta^{0}\|_{2}^{2}}{n} + \lambda \|\beta^{0}\|_{1}$$

We now replace  $\mathbf{Y}$  using equation (1.1).

$$\frac{\|\mathbf{Y} - \mathbf{X}\hat{\boldsymbol{\beta}}\|_{2}^{2}}{n} + \lambda \|\hat{\boldsymbol{\beta}}\|_{1} \leq \frac{\|\mathbf{Y} - \mathbf{X}\boldsymbol{\beta}^{0}\|_{2}^{2}}{n} + \lambda \|\boldsymbol{\beta}^{0}\|_{1}$$

$$\Rightarrow \frac{\|(\mathbf{X}\boldsymbol{\beta}^{0} + \boldsymbol{\varepsilon}) - \mathbf{X}\hat{\boldsymbol{\beta}}\|_{2}^{2}}{n} + \lambda \|\hat{\boldsymbol{\beta}}\|_{1} \leq \frac{\|(\mathbf{X}\boldsymbol{\beta}^{0} + \boldsymbol{\varepsilon}) - \mathbf{X}\boldsymbol{\beta}^{0}\|_{2}^{2}}{n} + \lambda \|\boldsymbol{\beta}^{0}\|_{1}$$

$$\Rightarrow \frac{\|\mathbf{X}(\boldsymbol{\beta}^{0} - \hat{\boldsymbol{\beta}}) + \boldsymbol{\varepsilon}\|_{2}^{2}}{n} + \lambda \|\hat{\boldsymbol{\beta}}\|_{1} \leq \frac{\|\mathbf{X}(\boldsymbol{\beta}^{0} - \boldsymbol{\beta}^{0}) + \boldsymbol{\varepsilon}\|_{2}^{2}}{n} + \lambda \|\boldsymbol{\beta}^{0}\|_{1}$$

$$\Rightarrow \frac{\|\mathbf{X}(\boldsymbol{\beta}^{0} - \hat{\boldsymbol{\beta}}) + \boldsymbol{\varepsilon}, \mathbf{X}(\boldsymbol{\beta}^{0} - \hat{\boldsymbol{\beta}}) + \boldsymbol{\varepsilon}\rangle}{n} + \lambda \|\hat{\boldsymbol{\beta}}\|_{1} \leq \frac{\|\boldsymbol{\varepsilon}\|_{2}^{2}}{n} + \lambda \|\boldsymbol{\beta}^{0}\|_{1}$$

$$\Rightarrow \frac{\|\mathbf{X}(\boldsymbol{\beta}^{0} - \hat{\boldsymbol{\beta}})\|_{2}^{2} + \|\boldsymbol{\varepsilon}\|_{2}^{2} + 2\langle\mathbf{X}(\boldsymbol{\beta}^{0} - \hat{\boldsymbol{\beta}}), \boldsymbol{\varepsilon}\rangle}{n} + \lambda \|\hat{\boldsymbol{\beta}}\|_{1} \leq \frac{\|\boldsymbol{\varepsilon}\|_{2}^{2}}{n} + \lambda \|\boldsymbol{\beta}^{0}\|_{1}$$

$$\Rightarrow \frac{\|\mathbf{X}(\hat{\boldsymbol{\beta}} - \boldsymbol{\beta}^{0})\|_{2}^{2}}{n} + \lambda \|\hat{\boldsymbol{\beta}}\|_{1} \leq \frac{2\langle\mathbf{X}(\hat{\boldsymbol{\beta}} - \boldsymbol{\beta}^{0}), \boldsymbol{\varepsilon}\rangle}{n} + \lambda \|\boldsymbol{\beta}^{0}\|_{1}$$

$$\Rightarrow \frac{\|\mathbf{X}(\hat{\boldsymbol{\beta}} - \boldsymbol{\beta}^{0})\|_{2}^{2}}{n} + \lambda \|\hat{\boldsymbol{\beta}}\|_{1} \leq \frac{2\boldsymbol{\varepsilon}^{T}\mathbf{X}(\hat{\boldsymbol{\beta}} - \boldsymbol{\beta}^{0})}{n} + \lambda \|\boldsymbol{\beta}^{0}\|_{1}$$

Let

$$\mathscr{T} := \left\{ \max_{1 \le j \le p} 2 \frac{\left| \varepsilon^T \mathbf{X}^{(j)} \right|}{n} \le \lambda_0 \right\}$$

**Lemma 1.2** (Lemma 6.2.). For all t > 0 and

$$\lambda_0 := 2\sigma \sqrt{\frac{t^2 + 2\log p}{n}}$$

we have

$$\mathbb{P}(\mathscr{T}) \geq 1 - 2 \exp\left[-t^2/2\right]$$

*Proof.* We define

$$V_j := \frac{\varepsilon^T \mathbf{X}^{(j)}}{\sqrt{n\sigma^2}}$$

1.1. SECTION 6.2

Then we have

$$\mathbb{P}(\mathscr{T}) = \mathbb{P}\left(\left\{\max_{1 \leq j \leq p} 2 \frac{\left|\varepsilon^{T} \mathbf{X}^{(j)}\right|}{n} \leq 2\sigma \sqrt{\frac{t^{2} + 2\log p}{n}}\right\}\right)$$

$$= \mathbb{P}\left(\left\{\max_{1 \leq j \leq p} \frac{\varepsilon^{T} \mathbf{X}^{(j)}}{\sqrt{n\sigma^{2}}} \leq \sqrt{t^{2} + 2\log p}\right\}\right)$$

$$= \mathbb{P}\left(\left\{\max_{1 \leq j \leq p} |V_{j}| \leq \sqrt{t^{2} + 2\log p}\right\}\right)$$

$$= 1 - \mathbb{P}\left(\left\{\max_{1 \leq j \leq p} |V_{j}| > \sqrt{t^{2} + 2\log p}\right\}\right)$$

**Lemma 1.3** (Lemma 6.3.). We have on  $\mathcal{T}$ , with  $\lambda \geq 2\lambda_0$ ,

$$2\frac{\|\mathbf{X}(\hat{\beta} - \beta^0)\|_2^2}{n} + \lambda \|\hat{\beta}_{S_0^c}\|_1 \le 3\lambda \|\hat{\beta}_{S_0} - \beta_{S_0}^0\|_1$$

*Proof.* We start with the Basic Inequality

$$\frac{\|\mathbf{X}(\hat{\beta} - \beta^0)\|_2^2}{n} + \lambda \|\hat{\beta}\|_1 \le 2\frac{\varepsilon^T \mathbf{X}(\hat{\beta} - \beta^0)}{n} + \lambda \|\beta^0\|_1$$

Now since we are on  $\mathcal{T}$  and since  $2\lambda_0 \leq \lambda$ 

$$\frac{\|\mathbf{X}(\hat{\beta} - \beta^0)\|_2^2}{n} + \lambda \|\hat{\beta}\|_1 \le \lambda_0 \|\hat{\beta} - \beta^0\|_1 + \lambda \|\beta^0\|_1$$

$$2\frac{\|\mathbf{X}(\hat{\beta} - \beta^0)\|_2^2}{n} + 2\lambda \|\hat{\beta}\|_1 \le \lambda \|\hat{\beta} - \beta^0\|_1 + 2\lambda \|\beta^0\|_1$$

By using the triangle inequality on the left hand side

$$\begin{split} \|\hat{\beta}\|_{1} &= \|\hat{\beta}_{S_{0}}\|_{1} + \|\hat{\beta}_{S_{0}^{c}}\|_{1} \\ &= \|\beta_{S_{0}}^{0} - \beta_{S_{0}}^{0} + \hat{\beta}_{S_{0}}\|_{1} + \|\hat{\beta}_{S_{0}^{c}}\|_{1} \\ &\geq \|\beta_{S_{0}}^{0}\|_{1} - \|\hat{\beta}_{S_{0}} - \beta_{S_{0}}^{0}\|_{1} + \|\hat{\beta}_{S_{0}^{c}}\|_{1} \end{split}$$

whereas on the right hand side

$$\|\hat{\beta} - \beta^{0}\|_{1} = \|(\hat{\beta}_{S_{0}} + \hat{\beta}_{S_{0}^{c}}) - (\beta_{S_{0}}^{0} + \underbrace{\beta_{S_{0}^{c}}^{0}}_{=0})\|_{1}$$
$$= \|\hat{\beta}_{S_{0}} - \beta_{S_{0}}^{0}\|_{1} + \|\hat{\beta}_{S_{0}^{c}}\|_{1}$$

Injecting these two results, we get that

$$2\frac{\|\mathbf{X}(\hat{\beta} - \beta^{0})\|_{2}^{2}}{n} + 2\lambda \|\hat{\beta}\|_{1} \leq \lambda \|\hat{\beta} - \beta^{0}\|_{1} + 2\lambda \|\beta^{0}\|_{1}$$

$$\Rightarrow 2\frac{\|\mathbf{X}(\hat{\beta} - \beta^{0})\|_{2}^{2}}{n} + 2\lambda \left(\|\beta_{S_{0}}^{0}\|_{1} - \|\hat{\beta}_{S_{0}} - \beta_{S_{0}}^{0}\|_{1} + \|\hat{\beta}_{S_{0}^{c}}\|_{1}\right)$$

$$\leq \lambda \left(\|\hat{\beta}_{S_{0}} - \beta_{S_{0}}^{0}\|_{1} + \|\hat{\beta}_{S_{0}^{c}}\|_{1}\right) + 2\lambda \|\beta^{0}\|_{1}$$

$$\Rightarrow 2\frac{\|\mathbf{X}(\hat{\beta} - \beta^{0})\|_{2}^{2}}{n} + 2\lambda \|\underbrace{\beta_{S_{0}^{0}}^{0}}_{=\beta^{0}}\|_{1} + \lambda \|\hat{\beta}_{S_{0}^{c}}\|_{1} \leq 3\lambda \|\hat{\beta}_{S_{0}} - \beta_{S_{0}}^{0}\|_{1} + 2\lambda \|\beta^{0}\|_{1}$$

$$\Rightarrow 2\frac{\|\mathbf{X}(\hat{\beta} - \beta^{0})\|_{2}^{2}}{n} + \lambda \|\hat{\beta}_{S_{0}^{c}}\|_{1} \leq 3\lambda \|\hat{\beta}_{S_{0}} - \beta_{S_{0}}^{0}\|_{1}$$

**Theorem 1.4** (Compatibility condition). We say that the compatibility condition is met for the set  $S_0$ , if for some  $\phi_0 > 0$ , and for all  $\beta$  satisfying  $\|\beta_{S_0^c}\|_1 \leq 3\|\beta_{S_0}\|_1$ , it holds that

 $\|\beta_{S_0}\|_1^2 \le \left(\beta^T \hat{\Sigma} \beta\right) s_0 / \phi_0^2$ 

**Theorem 1.5** (Theorem 6.1.). Suppose the compatibility condition holds for  $S_0$ . Then on  $\mathcal{T}$ , we have for  $\lambda \geq 2\lambda_0$ ,

$$\left\| \mathbf{X} \left( \hat{\beta} - \beta^0 \right) \right\|_2^2 / n + \lambda \left\| \hat{\beta} - \beta^0 \right\|_1 \le 4\lambda^2 s_0 / \phi_0^2$$

**Theorem 1.6** (Compatibility condition for general sets). We say that the compatibility condition holds for the set S, if for some constant  $\phi(S) > 0$ , and for all  $\beta$ , with  $\|\beta_{S^c}\|_1 \leq 3 \|\beta_S\|_1$ , one has

$$\|\beta_S\|_1^2 \le \left(\beta^T \hat{\Sigma} \beta\right) |S|/\phi^2(S)$$

#### 1.2 Section 6.3

Now  $\mathbb{E}[\mathbf{Y}] := \mathbf{f}^0$ 

**Lemma 1.7** (New version of the Basic Inequality).  $\forall \beta^* \in \mathbb{R}^p$  we have

$$\frac{\|\mathbf{X}\hat{\beta} - \mathbf{f}^{0}\|_{2}^{2}}{n} + \lambda \|\hat{\beta}\|_{1} \le \frac{\|\mathbf{X}\beta^{*} - \mathbf{f}^{0}\|_{2}^{2}}{n} + \lambda \|\beta^{*}\|_{1} + \frac{2\varepsilon^{T}\mathbf{X}(\hat{\beta} - \beta^{*})}{n}$$
(1.3)

1.2. SECTION 6.3

*Proof.* By definition of  $\hat{\beta}$ , we have that

$$\forall \beta \quad \frac{\|\mathbf{Y} - \mathbf{X}\hat{\beta}\|_2^2}{n} + \lambda \|\hat{\beta}\|_1 \le \frac{\|\mathbf{Y} - \mathbf{X}\beta\|_2^2}{n} + \lambda \|\beta\|_1$$

In particular for  $\beta = \beta^*$  we have

$$\forall \beta^* \quad \frac{\|\mathbf{Y} - \mathbf{X}\hat{\beta}\|_2^2}{n} + \lambda \|\hat{\beta}\|_1 \le \frac{\|\mathbf{Y} - \mathbf{X}\beta^*\|_2^2}{n} + \lambda \|\beta\|_1$$

We since  $\mathbf{Y} = \mathbf{f}^0 + \boldsymbol{\varepsilon}$ 

$$\frac{\|\mathbf{Y} - \mathbf{X}\hat{\boldsymbol{\beta}}\|_{2}^{2}}{n} + \lambda \|\hat{\boldsymbol{\beta}}\|_{1} \leq \frac{\|\mathbf{Y} - \mathbf{X}\boldsymbol{\beta}^{*}\|_{2}^{2}}{n} + \lambda \|\boldsymbol{\beta}^{0}\|_{1}$$

$$\Rightarrow \frac{\|(\mathbf{f}^{0} + \boldsymbol{\varepsilon}) - \mathbf{X}\hat{\boldsymbol{\beta}}\|_{2}^{2}}{n} + \lambda \|\hat{\boldsymbol{\beta}}\|_{1} \leq \frac{\|(\mathbf{f}^{0} + \boldsymbol{\varepsilon}) - \mathbf{X}\boldsymbol{\beta}^{*}\|_{2}^{2}}{n} + \lambda \|\boldsymbol{\beta}^{*}\|_{1}$$

$$\Rightarrow \frac{\|(\mathbf{f}^{0} - \mathbf{X}\hat{\boldsymbol{\beta}}) + \boldsymbol{\varepsilon}\|_{2}^{2}}{n} + \lambda \|\hat{\boldsymbol{\beta}}\|_{1} \leq \frac{\|(\mathbf{f}^{0} - \mathbf{X}\boldsymbol{\beta}^{*}) + \boldsymbol{\varepsilon}\|_{2}^{2}}{n} + \lambda \|\boldsymbol{\beta}^{*}\|_{1}$$

$$\Rightarrow \frac{\langle(\mathbf{f}^{0} - \mathbf{X}\hat{\boldsymbol{\beta}}) + \boldsymbol{\varepsilon}, (\mathbf{f}^{0} - \mathbf{X}\hat{\boldsymbol{\beta}}) + \boldsymbol{\varepsilon}\rangle}{n} + \lambda \|\hat{\boldsymbol{\beta}}\|_{1}$$

$$\leq \frac{\langle(\mathbf{f}^{0} - \mathbf{X}\boldsymbol{\beta}^{*}) + \boldsymbol{\varepsilon}, (\mathbf{f}^{0} - \mathbf{X}\boldsymbol{\beta}^{*}) + \boldsymbol{\varepsilon}\rangle}{n} + \lambda \|\hat{\boldsymbol{\beta}}\|_{1}$$

$$\Rightarrow \frac{\|\mathbf{f}^{0} - \mathbf{X}\hat{\boldsymbol{\beta}}\|_{2}^{2} + \|\boldsymbol{\varepsilon}\|_{2}^{2} + 2\langle\mathbf{f}^{0} - \mathbf{X}\hat{\boldsymbol{\beta}}, \boldsymbol{\varepsilon}\rangle}{n} + \lambda \|\boldsymbol{\beta}^{*}\|_{1}$$

$$\leq \frac{\|\mathbf{f}^{0} - \mathbf{X}\boldsymbol{\beta}^{*}\|_{2}^{2} + \|\boldsymbol{\varepsilon}\|_{2}^{2} + 2\langle\mathbf{f}^{0} - \mathbf{X}\boldsymbol{\beta}^{*}, \boldsymbol{\varepsilon}\rangle}{n} + \lambda \|\boldsymbol{\beta}^{*}\|_{1}$$

$$\Rightarrow \frac{\|\mathbf{X}\hat{\boldsymbol{\beta}} - \mathbf{f}^{0}\|_{2}^{2}}{n} + \lambda \|\hat{\boldsymbol{\beta}}\|_{1} \leq \frac{\|\mathbf{X}\boldsymbol{\beta}^{*} - \mathbf{f}^{0}\|_{2}^{2}}{n} + \lambda \|\boldsymbol{\beta}^{*}\|_{1} + \frac{2\langle\mathbf{X}(\hat{\boldsymbol{\beta}} - \boldsymbol{\beta}^{*}), \boldsymbol{\varepsilon}\rangle}{n}$$

$$\Rightarrow \frac{\|\mathbf{X}\hat{\boldsymbol{\beta}} - \mathbf{f}^{0}\|_{2}^{2}}{n} + \lambda \|\hat{\boldsymbol{\beta}}\|_{1} \leq \frac{\|\mathbf{X}\boldsymbol{\beta}^{*} - \mathbf{f}^{0}\|_{2}^{2}}{n} + \lambda \|\boldsymbol{\beta}^{*}\|_{1} + \frac{2\boldsymbol{\varepsilon}^{T}\mathbf{X}(\hat{\boldsymbol{\beta}} - \boldsymbol{\beta}^{*})}{n}$$

Lemma 1.8 (New version of the Lemma 6.3.).

$$\frac{4\|\mathbf{X}\hat{\beta} - \mathbf{f}^0\|_2^2}{n} + 3\lambda \|\hat{\beta}_{S_*^c}\|_1 \le 5\lambda \|\hat{\beta}_{S_*} - \beta_{S_*}^*\|_1 + \frac{4\|\mathbf{X}\beta^* - \mathbf{f}^0\|_2^2}{n}$$
(1.4)

# Chapter 2

## Introduction

- 2.1 The first section
- 2.1.1 The first subsection

To be added

- how to get  $\hat{b}$  on page 101.
- where the  $\chi^2$  distribution comes from in page 101