

# Realization on Linear Regression Model

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# Linear equations

We observe couples  $(y_1, x_1), (y_2, x_2), (y_3, x_3), \dots$ ,  
where  $y_t \in \mathbb{R}, x_t \in \mathbb{R}^d \quad \forall t \in \mathbb{N}$ .

We suppose that members of couples are connected by linear  
equations

$$y_t = x_t^\top \beta_0 + e_t \quad \forall t \in \mathbb{N},$$

where disturbances  $e_t \in \mathbb{R}$  are unknown and  $\beta_0 \in \Theta \subset \mathbb{R}^d$  is  
unknown parameter common for all equations.

Our aim is to reasonably approximate  $\beta_0$  from couples  $(y_t, x_t)$ ,  
 $t \in \{1, 2, \dots, T\}$ , where  $T \in \mathbb{N}$ .

Inaccuracy in any equation will be measured using a nonnegative Borel measurable function

$$\rho : \mathbb{R} \rightarrow \mathbb{R}, \rho \geq 0.$$

As approximation of  $\beta_0$ , we take a parameter  $\beta \in \Theta$  for which the corresponding disturbances are “small”.

# Linear equations

More precisely, we possess given positive numbers  $\varepsilon_T > 0$ ,  $T \in \mathbb{N}$  with

$$\limsup_{T \rightarrow +\infty} \varepsilon_T = \bar{\varepsilon}, \quad 0 \leq \bar{\varepsilon} < +\infty$$

and we find  $\hat{\beta}_T \in \Theta$  such that

$$\frac{1}{T} \sum_{t=1}^T \rho(y_t - x_t^\top \hat{\beta}_T) < \Delta_T + \varepsilon_T,$$

where

$$\Delta_T = \inf \left\{ \frac{1}{T} \sum_{t=1}^T \rho(y_t - x_t^\top \beta) : \beta \in \Theta \right\}.$$

# Functional setting

The problem can be generalized using a function defined for all  $\beta \in \Theta$  and for all  $\mu$  probability Borel measures on  $\mathbb{R}^{d+1}$  by the formula

$$f(\beta; \mu) = \int \rho(\mathbf{x}^\top (\beta_0 - \beta) + \mathbf{e}) \mu(d\mathbf{x}, d\mathbf{e}).$$

The integral is always well-defined since  $\rho$  is a nonnegative Borel measurable function and  $\mu$  is a probability Borel measures.

Let us consider a sequence  $\mu_T$ ,  $T \in \mathbb{N}$  of probability Borel measures on  $\mathbb{R}^{d+1}$ .

We find a parameter  $\hat{\beta}_T \in \Theta$  such that

$$\int \rho\left(\mathbf{e} + \mathbf{x}^\top (\beta_0 - \hat{\beta}_T)\right) \mu_T(d\mathbf{x}, d\mathbf{e}) < \Delta_T + \varepsilon_T,$$

where

$$\Delta_T = \inf \left\{ \int \rho(\mathbf{x}^\top (\beta_0 - \beta) + \mathbf{e}) \mu_T(d\mathbf{x}, d\mathbf{e}) : \beta \in \Theta \right\}.$$

Setting

$$\mu_T = \sum_{t=1}^T \frac{1}{T} \delta_{(y_t, e_t)}$$

we receive precisely the original problem

$$f(\beta; \mu_T) = \frac{1}{T} \sum_{t=1}^T \rho(e_t + x_t^\top (\beta_0 - \beta)) = \frac{1}{T} \sum_{t=1}^T \rho(y_t - x_t^\top \beta).$$

**LEMMA:**

For any  $\nu$  probability Borel measures on  $\mathbb{R}^{d+1}$  and any  $\Delta > 0$  we have

$$\begin{aligned} & \lim_{\kappa \rightarrow +\infty} \inf_{\|\gamma\|=1} \nu(\{(x, e) : \kappa |x^\top \gamma| \geq \Delta + |e|\}) = \\ & = \inf \{ \nu(\{(x, e) \in \mathbb{R}^{d+1} : x^\top \gamma \neq 0\}) : \|\gamma\| = 1, \gamma \in \mathbb{R}^d \}. \end{aligned}$$



**PROOF:**

Let us denote

$$M = \inf \{ \nu(\{(x, e) \in \mathbb{R}^{d+1} : x^T \gamma \neq 0\}) : \|\gamma\| = 1, \gamma \in \mathbb{R}^d \}.$$

1)

Let  $\gamma \in \mathbb{R}^d$  and  $k \in \mathbb{N}$ , then

$$\{(x, e) : k |x^T \gamma| \geq \Delta + |e|\} \subset \{(x, e) : x^T \gamma \neq 0\}.$$

Therefore,

$$\lim_{\kappa \rightarrow +\infty} \inf_{\|\gamma\|=1} \nu(\{(x, e) : \kappa |x^T \gamma| \geq \Delta + |e|\}) \leq M.$$

2)

Consider a sequence  $\gamma_k \in \mathbb{R}^d$ ,  $\|\gamma_k\| = 1$  for all  $k \in \mathbb{N}$  such that

$$\begin{aligned} \nu(\{(x, e) : k |x^\top \gamma_k| \geq \Delta + |e|\}) &< \\ &< \inf_{\|\gamma\|=1} \nu(\{(x, e) : k |x^\top \gamma| \geq \Delta + |e|\}) + \frac{1}{k}. \end{aligned}$$

The sequence belongs to a compact, therefore, it contains a convergent subsequence

$$\lim_{j \rightarrow +\infty} \gamma_{k_j} = \hat{\gamma}, \quad \|\hat{\gamma}\| = 1.$$

Hence,

$$\begin{aligned} & \{(\mathbf{x}, \mathbf{e}) : \mathbf{x}^\top \hat{\gamma} \neq 0\} \\ &= \bigcup_{J=1}^{+\infty} \bigcap_{j=J}^{+\infty} \left\{ (\mathbf{x}, \mathbf{e}) : |\mathbf{x}^\top \hat{\gamma}| \geq |\mathbf{x}^\top (\hat{\gamma} - \gamma_{k_j})| + \frac{1}{k_j} (\Delta + |\mathbf{e}|) \right\} \\ &\subset \bigcup_{J=1}^{+\infty} \bigcap_{j=J}^{+\infty} \{(\mathbf{x}, \mathbf{e}) : k_j |\mathbf{x}^\top \gamma_{k_j}| \geq \Delta + |\mathbf{e}|\}. \end{aligned}$$

Because of  $\sigma$ -additivity of the measure  $\nu$ , we have

$$\begin{aligned}
 M &\leq \nu(\{(x, e) : x^T \hat{\gamma} \neq 0\}) \\
 &\leq \lim_{J \rightarrow +\infty} \nu\left(\bigcap_{j=J}^{+\infty} \{(x, e) : k_j |x^T \gamma_{k_j}| \geq \Delta + |e|\}\right) \\
 &\leq \liminf_{J \rightarrow +\infty} \nu(\{(x, e) : k_J |x^T \gamma_{k_J}| \geq \Delta + |e|\}) \\
 &\leq \liminf_{J \rightarrow +\infty} \left[ \inf_{\|\gamma\|=1} \nu(\{(x, e) : \kappa |x^T \gamma| \geq \Delta + |e|\}) + \frac{1}{k_J} \right] \\
 &= \inf_{\|\gamma\|=1} \nu(\{(x, e) : \kappa |x^T \gamma| \geq \Delta + |e|\}).
 \end{aligned}$$

Q.E.D.

**Assumption 1:**  $\Theta \subset \mathbb{R}^d$  is a closed set and for each  $\beta \in \Theta$

$$\int \rho(\mathbf{e}) \nu(\mathbf{d}\mathbf{x}, \mathbf{d}\mathbf{e}) \leq \int \rho(\mathbf{e} + \mathbf{x}^\top(\beta_0 - \beta)) \nu(\mathbf{d}\mathbf{x}, \mathbf{d}\mathbf{e}).$$

**Assumption 2:** There is a function  $\psi : \mathbb{R}_+ \rightarrow \mathbb{R}_+$  which is continuous, nondecreasing and fulfilling:

1. For all  $t \in \mathbb{R}$   $\rho(t) \leq \psi(|t|)$ .
2. For all  $t > 0$   $\int \psi(|\mathbf{e}| + t\|\mathbf{x}\|) \nu(\mathbf{d}\mathbf{x}, \mathbf{d}\mathbf{e}) < +\infty$ .
3. For all  $t > 0$   $\int \psi(|\mathbf{e}| + t\|\mathbf{x}\|) \mu_T(\mathbf{d}\mathbf{x}, \mathbf{d}\mathbf{e}) < +\infty$ .

## Assumption 3:

$$\mu_T \xrightarrow[T \rightarrow +\infty]{\psi - \mathcal{D}} \nu$$

which means

$\forall f : \mathbb{R}^{d+1} \rightarrow \mathbb{R}$  bounded continuous

$$\int f(x, e) \mu_T(dx, de) \xrightarrow[T \rightarrow +\infty]{} \int f(x, e) \nu(dx, de),$$

$$\forall t > 0 \int \psi(|e| + t\|x\|) \mu_T(dx, de) \xrightarrow[T \rightarrow +\infty]{} \int \psi(|e| + t\|x\|) \nu(dx, de).$$

# Assumptions

Assumption 4: Denoting

$$H_\rho = \liminf_{\Delta \rightarrow +\infty} \inf \{ \rho(t) : |t| > \Delta, t \in \mathbb{R} \},$$

$$M = \inf \{ \nu(\{(x, e) \in \mathbb{R}^{d+1} : x^\top \gamma \neq 0\}) : \|\gamma\| = 1, \gamma \in \mathbb{R}^d \},$$

we require  $M > 0$  and a balance

$$H_\rho M > \int \rho(e) \nu(dx, de) + \limsup_{T \rightarrow +\infty} \varepsilon_T.$$

# Notation

We denote

$$g(\beta) = \int \rho(\mathbf{e} + \mathbf{x}^\top(\beta_0 - \beta)) \nu(d\mathbf{x}, d\mathbf{e}) \quad \forall \beta \in \Theta.$$

According to Assumption 1, we know that  $\beta_0$  is a minimizer of  $g$ .

The set of all minimizers and the set of all  $\varepsilon$ -minimizers of  $g$  are denoted by

$$\begin{aligned}\Phi(g) &= \{\beta \in \Theta : g(\beta) = g(\beta_0)\}, \\ \Psi(g; \varepsilon) &= \{\beta \in \Theta : g(\beta) \leq g(\beta_0) + \varepsilon\}.\end{aligned}$$

We recall that we possess given positive numbers  $\varepsilon_T > 0$ ,  $T \in \mathbb{N}$  with

$$\limsup_{T \rightarrow +\infty} \varepsilon_T = \bar{\varepsilon}, \quad 0 \leq \bar{\varepsilon} < +\infty.$$



**THEOREM:**

Let the previous **Assumptions 1-4** be fulfilled.

Then  $\beta_0 \in \Phi \langle \mathbf{g} \rangle$  and  $\hat{\beta}_T$  exists for every  $T \in \mathbb{N}$ .

The sequence  $\hat{\beta}_T, T \in \mathbb{N}$  is **compact** and

$$\emptyset \neq \text{Ls} \left\{ \hat{\beta}_T, T \in \mathbb{N} \right\} \subset \Psi \langle \mathbf{g}; \bar{\varepsilon} \rangle,$$

$$\lim_{T \rightarrow +\infty} d \left( \hat{\beta}_T, \Psi \langle \mathbf{g}; \bar{\varepsilon} \rangle \right) = 0.$$

**PROOF:**

1)

We know that for each  $\beta \in \Theta$ ,  $T \in \mathbb{N}$

$$f(\beta; \mu_T), f(\beta; \nu) \in \mathbb{R}.$$

2)

Let  $\beta, \beta_T \in \Theta$  such that  $\beta_T \xrightarrow{T \rightarrow +\infty} \beta$ .

Let us fix  $\varepsilon > 0$ .

Then there exist  $\Delta, Q$  and a compact  $\bar{K} \subset \mathbb{R}^{d+1}$  such that

$$\|\beta_T - \beta_0\| \leq \Delta \quad \text{for all } T \in \mathbb{N},$$

$$\int_{\psi(|e| + \Delta\|x\|) > Q} (\psi(|e| + \Delta\|x\|) - Q) \nu(dx, de) < \varepsilon,$$

$$\mu_T(\mathbb{R}^{d+1} \setminus \bar{K}) < \frac{\varepsilon}{Q} \quad \text{for all } T \in \mathbb{N}.$$

Now, we are able to derive a convergence.

$$\begin{aligned}
 f(\beta_T; \mu_T) &= \int \rho(\mathbf{e} + \mathbf{x}^\top (\beta_0 - \beta_T)) \mu_T(\mathbf{d}\mathbf{x}, \mathbf{d}\mathbf{e}) = \\
 &= \int \min\{Q, \rho(\mathbf{x}^\top (\beta_0 - \beta) + \mathbf{e})\} \mu_T(\mathbf{d}\mathbf{x}, \mathbf{d}\mathbf{e}) + \\
 &+ \int \min\{Q, \rho(\mathbf{e} + \mathbf{x}^\top (\beta_0 - \beta_T))\} - \min\{Q, \rho(\mathbf{x}^\top (\beta_0 - \beta) + \mathbf{e})\} \mu_T(\mathbf{d}\mathbf{x}, \mathbf{d}\mathbf{e}) \\
 &+ \int_{\rho(\mathbf{e} + \mathbf{x}^\top (\beta_0 - \beta_T)) > Q} (\rho(\mathbf{e} + \mathbf{x}^\top (\beta_0 - \beta_T)) - Q) \mu_T(\mathbf{d}\mathbf{x}, \mathbf{d}\mathbf{e}).
 \end{aligned}$$

The function  $\min\{Q, \rho\}$  is bounded and continuous. Therefore,

$$\int \min\{Q, \rho(\mathbf{x}^\top(\beta_0 - \beta) + \mathbf{e})\} \mu_T(\mathbf{d}\mathbf{x}, \mathbf{d}\mathbf{e}) \xrightarrow{T \rightarrow +\infty} \int \min\{Q, \rho(\mathbf{x}^\top(\beta_0 - \beta) + \mathbf{e})\} \nu(\mathbf{d}\mathbf{x}, \mathbf{d}\mathbf{e}).$$

The second term fulfills

$$\begin{aligned}
 & \left| \int \min\{Q, \rho(\mathbf{e} + \mathbf{x}^\top (\beta_0 - \beta_T))\} - \min\{Q, \rho(\mathbf{x}^\top (\beta_0 - \beta) + \mathbf{e})\} \mu_T(d\mathbf{x}, d\mathbf{e}) \right| \\
 & < 2\varepsilon + \left| \int_{\bar{K}} \min\{Q, \rho(\mathbf{e} + \mathbf{x}^\top (\beta_0 - \beta_T))\} - \min\{Q, \rho(\mathbf{x}^\top (\beta_0 - \beta) + \mathbf{e})\} \mu_T(d\mathbf{x}, d\mathbf{e}) \right| \\
 & \leq 2\varepsilon + \sup_{(\mathbf{x}, \mathbf{e}) \in \bar{K}} \left| \min\{Q, \rho(\mathbf{e} + \mathbf{x}^\top (\beta_0 - \beta_T))\} - \min\{Q, \rho(\mathbf{x}^\top (\beta_0 - \beta) + \mathbf{e})\} \right| \\
 & \xrightarrow{T \rightarrow +\infty} 2\varepsilon
 \end{aligned}$$

because  $\rho$  is continuous, and, hence, uniformly continuous on each compact set.

## 3. term

The third term is smaller than  $\varepsilon$  since

$$\begin{aligned}
 0 &\leq \int_{\rho(\mathbf{e} + \mathbf{x}^\top (\beta_0 - \beta_T)) > Q} (\rho(\mathbf{e} + \mathbf{x}^\top (\beta_0 - \beta_T)) - Q) \mu_T(d\mathbf{x}, d\mathbf{e}) \\
 &\leq \int_{\psi(|\mathbf{e}| + T\|\mathbf{x}\|) > Q} (\psi(|\mathbf{e}| + T\|\mathbf{x}\|) - Q) \mu_T(d\mathbf{x}, d\mathbf{e}) \\
 &= \int \psi(|\mathbf{e}| + T\|\mathbf{x}\|) \mu_T(d\mathbf{x}, d\mathbf{e}) - \\
 &\quad - \int \min\{Q, \psi(|\mathbf{e}| + T\|\mathbf{x}\|)\} \mu_T(d\mathbf{x}, d\mathbf{e}) \\
 &\quad \xrightarrow{T \rightarrow +\infty} \\
 &\int_{\psi(|\mathbf{e}| + T\|\mathbf{x}\|) > Q} (\psi(|\mathbf{e}| + T\|\mathbf{x}\|) - Q) \nu(d\mathbf{x}, d\mathbf{e}) < \varepsilon.
 \end{aligned}$$

Thus, we proved

$$\lim_{T \rightarrow +\infty} f(\beta_T; \mu_T) = \int \rho(\mathbf{x}^\top (\beta_0 - \beta) + \mathbf{e}) \nu(d\mathbf{x}, d\mathbf{e}) = f(\beta; \nu).$$

It means that  $f(\beta; \mu_T)$  converge to  $f(\beta; \nu)$  uniformly on each compact.



# Proof - tightness

According to Assumptions, there is a number  $\Delta > 0$  such that

$$\inf_{|t| > \Delta} \rho(t) \cdot M > f(\beta_0; \nu) + \bar{\varepsilon}.$$

Hence according to Lemma, we are able to find  $\Gamma$  such that

$$\inf_{|t| > \Delta} \rho(t) \cdot \inf_{\|\gamma\|=1} \nu(\{(x, e) : \Gamma |x^\top \gamma| \geq \Delta + |e|\}) > f(\beta_0; \nu) + \bar{\varepsilon}.$$

Then, we define a compact

$$K = \{\beta \in \Theta : \|\beta - \beta_0\| \leq \Gamma\}.$$

For  $\beta \in \Theta$ ,  $\|\beta - \beta_0\| > \Gamma$  we receive following chain of inequalities:

$$\begin{aligned} f(\beta; \mu_T) &= \int \rho(\mathbf{x}^\top (\beta_0 - \beta) + \mathbf{e}) \mu_T(d\mathbf{x}, d\mathbf{e}) \\ &\geq \int_{|\mathbf{x}^\top (\beta_0 - \beta) + \mathbf{e}| > \Delta} \rho(\mathbf{x}^\top (\beta_0 - \beta) + \mathbf{e}) \mu_T(d\mathbf{x}, d\mathbf{e}) \\ &\geq \inf_{|t| > \Delta} \rho(t) \cdot \mu_T(\{(\mathbf{x}, \mathbf{e}) : |\mathbf{x}^\top (\beta_0 - \beta) + \mathbf{e}| > \Delta\}) \\ &\geq \inf_{|t| > \Delta} \rho(t) \cdot \mu_T(\{(\mathbf{x}, \mathbf{e}) : |\mathbf{x}^\top (\beta - \beta_0)| > \Delta + |\mathbf{e}|\}) \\ &\geq \inf_{|t| > \Delta} \rho(t) \cdot \mu_T\left(\left\{(\mathbf{x}, \mathbf{e}) : \Gamma \left| \mathbf{x}^\top \frac{\beta - \beta_0}{\|\beta - \beta_0\|} \right| > \Delta + |\mathbf{e}|\right\}\right). \end{aligned}$$

# Proof - tightness

For any  $\varepsilon > 0$ , properly chosen sequence of  $\gamma_T$ ,  $\|\gamma_T\| = 1$  and its cluster point  $\hat{\gamma}$ , we have








$$\begin{aligned}
 & \liminf_{T \rightarrow +\infty} \inf_{\beta \notin K} f(\beta; \mu_T) \geq \\
 & \geq \inf_{|t| > \Delta} \rho(t) \cdot \liminf_{T \rightarrow +\infty} \inf \{ \mu_T (\{(x, e) : \Gamma |x^T \gamma| > \Delta + |e|\}) : \|\gamma\| = 1 \} \\
 & \geq \inf_{|t| > \Delta} \rho(t) \cdot \liminf_{T \rightarrow +\infty} \mu_T (\{(x, e) : \Gamma |x^T \gamma_T| > \Delta + |e|\}) \\
 & \geq \inf_{|t| > \Delta} \rho(t) \cdot \liminf_{T \rightarrow +\infty} \mu_T (\{(x, e) : \Gamma |x^T \hat{\gamma}| > \Delta + |e| + \Gamma |x^T (\gamma_T - \hat{\gamma})|\}) \\
 & \geq \inf_{|t| > \Delta} \rho(t) \cdot \liminf_{T \rightarrow +\infty} \mu_T (\{(x, e) : \Gamma |x^T \hat{\gamma}| > (1 + \varepsilon)\Delta + |e|\}) \\
 & \geq \inf_{|t| > \Delta} \rho(t) \cdot \nu (\{(x, e) : \Gamma |x^T \hat{\gamma}| > (1 + \varepsilon)\Delta + |e|\}) .
 \end{aligned}$$







Letting  $\varepsilon$  vanish we have






$$\begin{aligned} \liminf_{T \rightarrow +\infty} \inf_{\beta \notin K} f(\beta; \mu_T) &\geq \inf_{|t| > \Delta} \rho(t) \cdot \nu(\{(x, e) : \Gamma |x^\top \hat{\gamma}| \geq \Delta + |e|\}) \\ &> f(\beta_0; \nu) + \bar{\varepsilon}. \end{aligned}$$

We found that it is sufficient to consider  $\beta \in K$ , only.  
We know that  $f(\beta; \mu_T)$  converge to  $f(\beta; \nu)$  uniformly on  $K$ .  
That completes the proof.

Q.E.D.

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