

# Robust Linear Regression by Subquantile Minimization

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# Presentation Overview

## ① Robust Regression

- Huber Contamination Model
- Blocks
- Preliminaries

## ② Empirical Results

- Table
- Figure

## ③ Mathematics

## ④ Referencing

# Huber Contamination Model

## Problem

The *Huber Contamination Model* is the following:

$$\hat{P} = (1 - \varepsilon)P + \varepsilon Q \text{ where } \varepsilon \in (0, 0.5)$$

where  $P$  and  $Q$  represent the general linear models

$$\mathbf{y}_P = \boldsymbol{\beta}_P^\top \mathbf{P} + \epsilon_P$$

$$\mathbf{y}_Q = \boldsymbol{\beta}_Q^\top \mathbf{Q} + \epsilon_Q$$

$\boldsymbol{\beta}_P$  and  $\boldsymbol{\beta}_Q$  are oracle regressors and  $\epsilon_P$  and  $\epsilon_Q$  represent 0-centered gaussian noise.

Our goal is to learn a model that learns a good distribution of  $P$  from  $\hat{P}$

# Motivation

## Definition

Oblivious Noise is noise which is independent of the input data

Adaptive Noise is noise which is dependent upon the input data

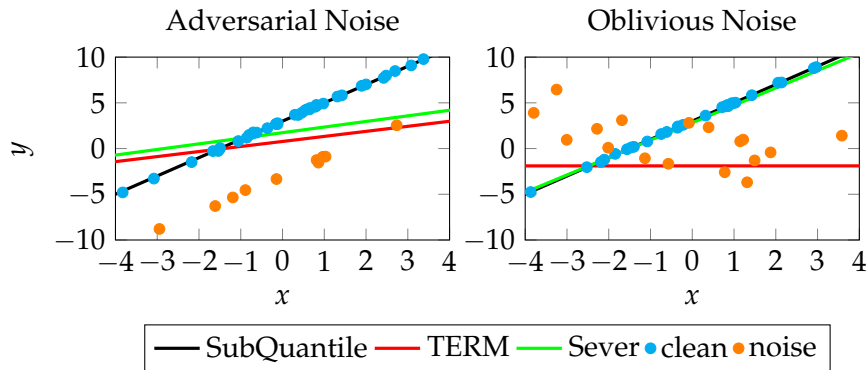


Figure: Sub-Quantile Performance on Adaptive Outliers

## Theorem

The expected optimal parameters of the corrupted model  $\hat{P}$

$$\mathbb{E} \left[ \mathbf{X}^\dagger \mathbf{y} \right] = (1 - \varepsilon) \boldsymbol{\beta}_P + \varepsilon \boldsymbol{\beta}_Q$$

where  $\mathbf{X}^\dagger \triangleq (\mathbf{X}^\top \mathbf{X})^{-1} \mathbf{X}^\top$ , i.e. the Moore-Penrose Inverse.

This theorem motivates our reasoning for optimizing over the Subquantile. We want a method to reduce  $\varepsilon$ .

# Statistical Preliminaries of the Subquantile

①

$$\mathcal{Q}_p = \inf \{x \in \mathbb{R} : p \leq F(x)\}$$

② Let  $\ell$  be the loss functions. We can now define risk as:

$$\mathcal{U} = \mathbb{E} [\ell(f(\mathbf{x}; \boldsymbol{\theta}), y)]$$

③ The  $p$ -Quantile of the Empirical Risk is given by:

$$\mathbb{L}_p = \frac{1}{p} \int_0^p \mathcal{Q}_q(\mathcal{U}) dq = \mathbb{E} [\mathcal{U} | \mathcal{U} \leq \mathcal{Q}_p(\mathcal{U})] = \max_{t \in \mathbb{R}} \left\{ t - \frac{1}{p} \mathbb{E} [(t - \mathcal{U})^+] \right\}$$

# Multiple Columns

Subtitle

## Heading

- ① Statement
- ② Explanation
- ③ Example

Lorem ipsum dolor sit amet, consectetur adipiscing elit. Integer lectus nisl, ultricies in feugiat rutrum, porttitor sit amet augue. Aliquam ut tortor mauris. Sed volutpat ante purus, quis accumsan dolor.

Objectives	Test RMSE (Drug Discovery)			
	$\epsilon = 0.1$	$\epsilon = 0.2$	$\epsilon = 0.3$	$\epsilon = 0.4$
ERM	1.303 <sub>(0.0665)</sub>	1.790 <sub>(0.0849)</sub>	2.198 <sub>(0.0645)</sub>	2.623 <sub>(0.1010)</sub>
CRR [?]	<b>1.079</b> <sub>(0.0899)</sub>	1.125 <sub>(0.0832)</sub>	1.385 <sub>(0.1372)</sub>	1.725 <sub>(0.1136)</sub>
STIR [?]	1.087 <sub>(0.1256)</sub>	1.167 <sub>(0.0750)</sub>	1.403 <sub>(0.0987)</sub>	1.668 <sub>(0.1142)</sub>
Robust Risk [?]	1.176 <sub>(0.1110)</sub>	1.336 <sub>(0.1882)</sub>	1.437 <sub>(0.1723)</sub>	1.800 <sub>(0.0820)</sub>
SMART [?]	1.094 <sub>(0.1065)</sub>	1.323 <sub>(0.0758)</sub>	1.578 <sub>(0.0799)</sub>	1.984 <sub>(0.2020)</sub>
TERM [?]	1.326 <sub>(0.0757)</sub>	1.357 <sub>(0.0990)</sub>	1.310 <sub>(0.0670)</sub>	1.302 <sub>(0.0851)</sub>
SEVER [?]	1.111 <sub>(0.0924)</sub>	<b>1.067</b> <sub>(0.0457)</sub>	<b>1.071</b> <sub>(0.0807)</sub>	<b>1.138</b> <sub>(0.1162)</sub>
Huber [?]	1.412 <sub>(0.0474)</sub>	1.501 <sub>(0.2918)</sub>	2.231 <sub>(0.9054)</sub>	2.247 <sub>(1.0399)</sub>
RANSAC [?]	1.238 <sub>(0.0529)</sub>	1.643 <sub>(0.1331)</sub>	2.092 <sub>(0.1935)</sub>	2.679 <sub>(0.1365)</sub>
SubQuantile( $p = 1 - \epsilon$ )	<b>0.887</b> <sub>(0.1046)</sub>	<b>0.936</b> <sub>(0.1051)</sub>	<b>0.927</b> <sub>(0.0729)</sub>	<b>1.015</b> <sub>(0.0978)</sub>
Genie ERM	0.986 <sub>(0.1039)</sub>	0.955 <sub>(0.0698)</sub>	1.038 <sub>(0.0886)</sub>	1.030 <sub>(0.0578)</sub>

**Table:** Drug Discovery Dataset. Empirical Risk over  $P$  with oblivious noise





# Rensselaer

# Definitions & Examples

## Definition

A **prime number** is a number that has exactly two divisors.

## Example

- 2 is prime (two divisors: 1 and 2).
- 3 is prime (two divisors: 1 and 3).
- 4 is not prime (**three** divisors: 1, 2, and 4).

You can also use the theorem, lemma, proof and corollary environments.

# Theorem, Corollary & Proof

Theorem (Mass–energy equivalence)

$$E = mc^2$$

Corollary

$$x + y = y + x$$

Proof.

$$\omega + \phi = \epsilon$$



# Equation

$$\cos^3 \theta = \frac{1}{4} \cos \theta + \frac{3}{4} \cos 3\theta \quad (1)$$

## Example (Theorem Slide Code)

```
\begin{frame}  
\frametitle{Theorem}  
\begin{theorem}[Mass--energy equivalence]  
$E = mc^2$  
\end{theorem}  
\end{frame}
```

Slide without title.

An example of the `\cite` command to cite within the presentation:

This statement requires citation [Smith, 2022, Kennedy, 2023].

# References



John Smith (2022)

Publication title

*Journal Name* 12(3), 45 – 678.



Annabelle Kennedy (2023)

Publication title

*Journal Name* 12(3), 45 – 678.



# Acknowledgements

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# The End

Questions? Comments?