



# Truthful Linear Regression

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$$y_i = \theta^\top x_i + z_i$$

$\hat{\theta}?$



$(x_1, y_1)$



$(x_2, y_2)$

⋮



$(x_n, y_n)$

$$y_i = \theta^\top x_i + z_i$$

$\hat{\theta}?$



$(x_1, y_1)$

$(x_2, y_2)$

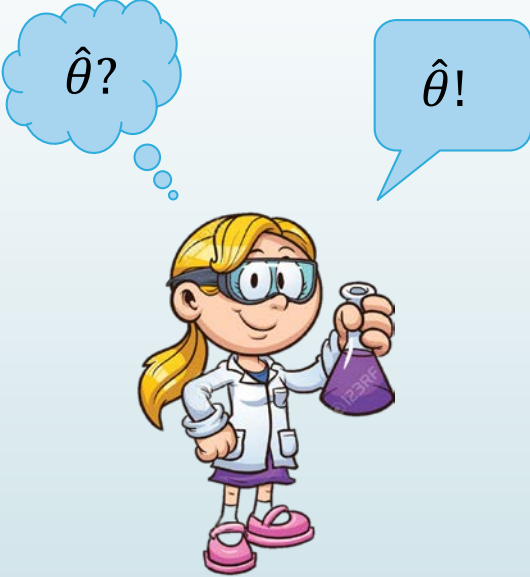
$(x_n, y_n)$



⋮



$$y_i = \theta^\top x_i + z_i$$



$(x_1, y_1)$

$(x_2, y_2)$

$(x_n, y_n)$



⋮



$$y_i = \theta^\top x_i + z_i$$

$\hat{\theta}?$



$(x_1, y_1)$



$(x_2, y_2)$

⋮



$(x_n, y_n)$

$$y_i = \theta^T x_i + z_i$$

$\hat{\theta}?$



$(x_1, y_1)$



$(x_2, y_2)$

⋮



$(x_n, y_n)$

$$y_i = \theta^\top x_i + z_i$$

$\hat{\theta}$ ? Privacy!



$$y_i = \theta^\top x_i + z_i$$

$\hat{\theta}?$

Privacy!



$(x_1, y_1)$

$(x_2, y_2)$

$(x_n, y_n)$





$$y_i = \theta^\top x_i + z_i$$

$\hat{\theta}?$

Privacy!

$\hat{\theta}!$

$(x_1, y_1)$

$(x_2, y_2)$

$(x_n, y_n)$



$$y_i = \theta^T x_i + z_i$$

$\hat{\theta}?$

Privacy!

$\hat{\theta}!$

$(x_1, y_1)$

$(x_2, y_2)$

$(x_n, y_n)$



- Challenges:
- Privacy guarantee
  - Data is not verifiable

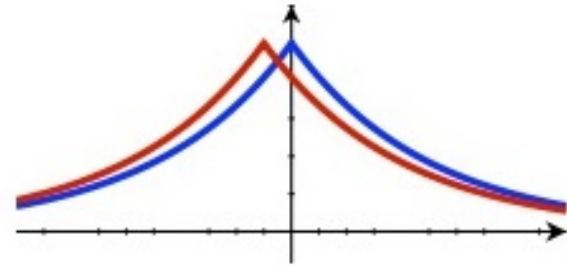
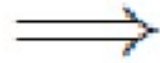


# Privacy

↓ A

$(t_1, \dots, t_i, \dots, t_n)$

$(t_1, \dots, t'_i, \dots, t_n)$



↓

↓

↓

↓

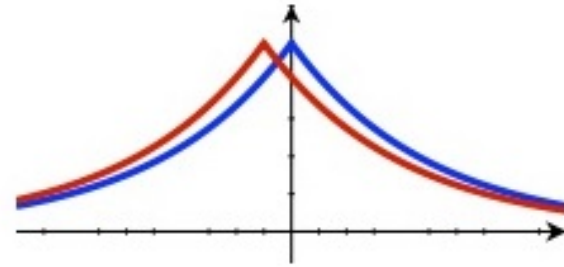
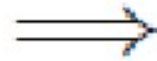
# Privacy

↓  $A$  is biased estimator

↓

$(t_1, \dots, t_i, \dots, t_n)$

$(t_1, \dots, t'_i, \dots, t_n)$



↓

➤ Getting non-trivial privacy parameter requires regularization

➤ Ridge regression

➤  $\theta$  is biased estimator

# Truthfulness

↓ Peer prediction - Ask



\_\_\_\_\_

# Truthfulness

↓ Peer prediction - Ask  $\theta$  is biased

↓

↓

↓

➡ Technical challenge:  $\theta$  is biased

Biased!

# Main result

We design a mechanism that is:

- ▶ private
- ▶ truthful
- ▶ accurately estimates  $\hat{\theta}$
- ▶ individually rational
- ▶ analyst's budget is asymptotically small

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