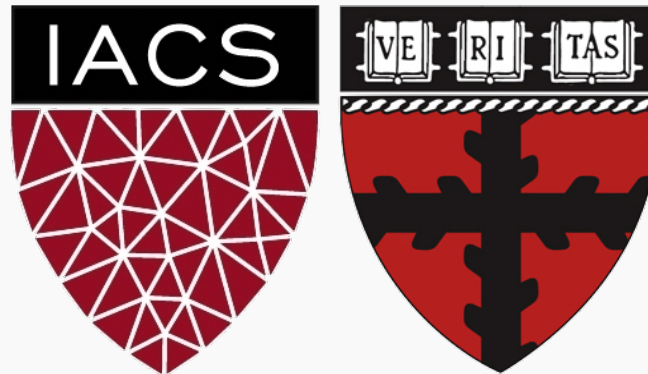


Prediction Intervals

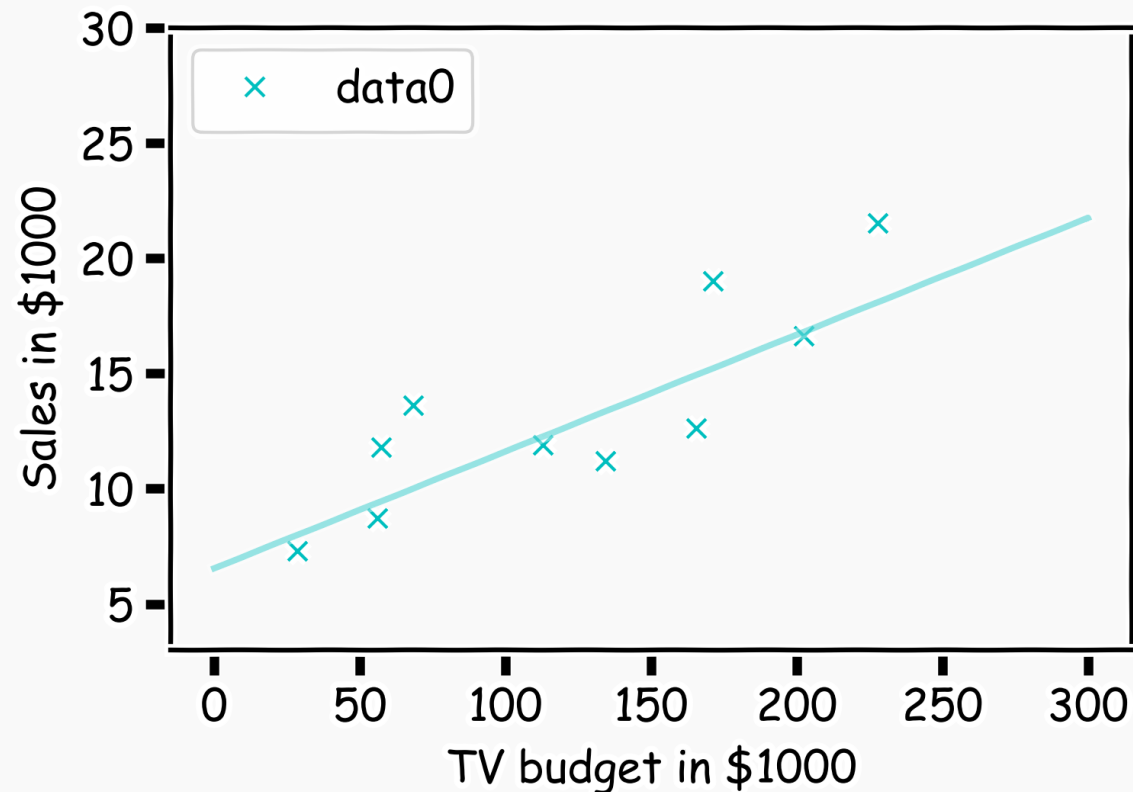
CS109A Introduction to Data Science

Pavlos Protopapas, Kevin Rader and Chris Tanner



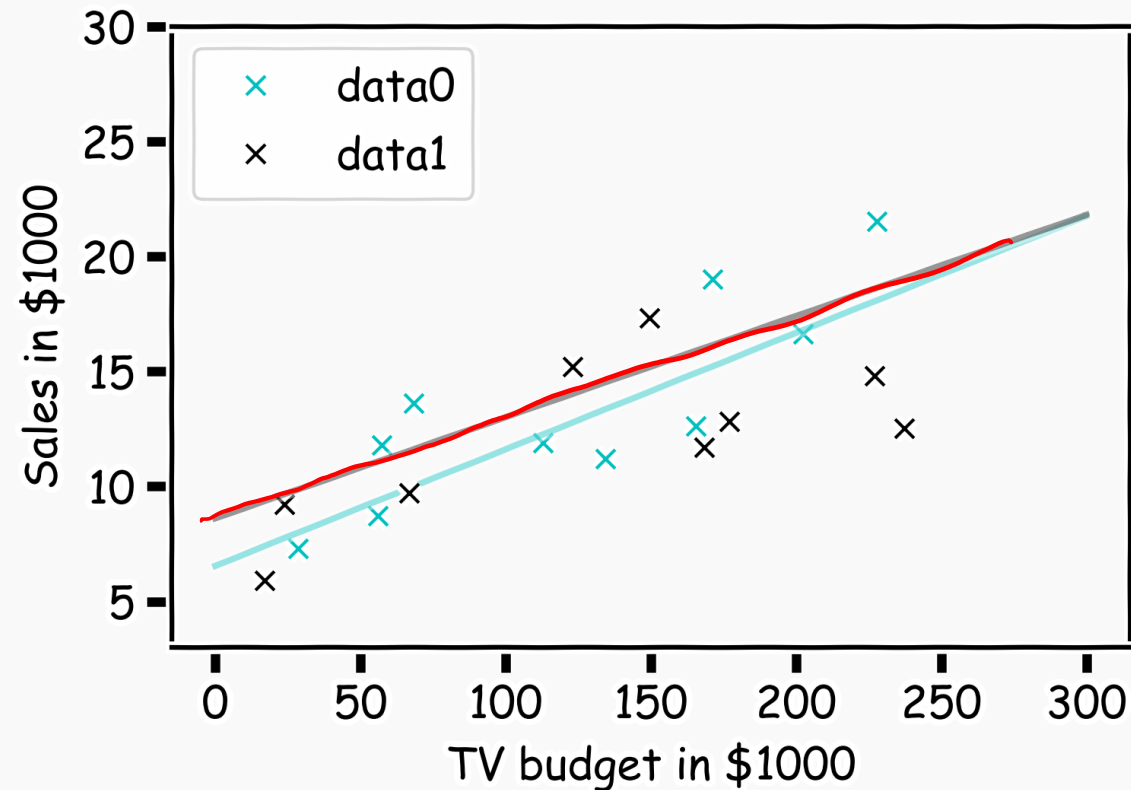
How well do we know \hat{f} ?

Our confidence in f is directly connected with our confidence in β s. For each bootstrap sample, we have one β , which we can use to determine the model, $f(x) = X\beta$.



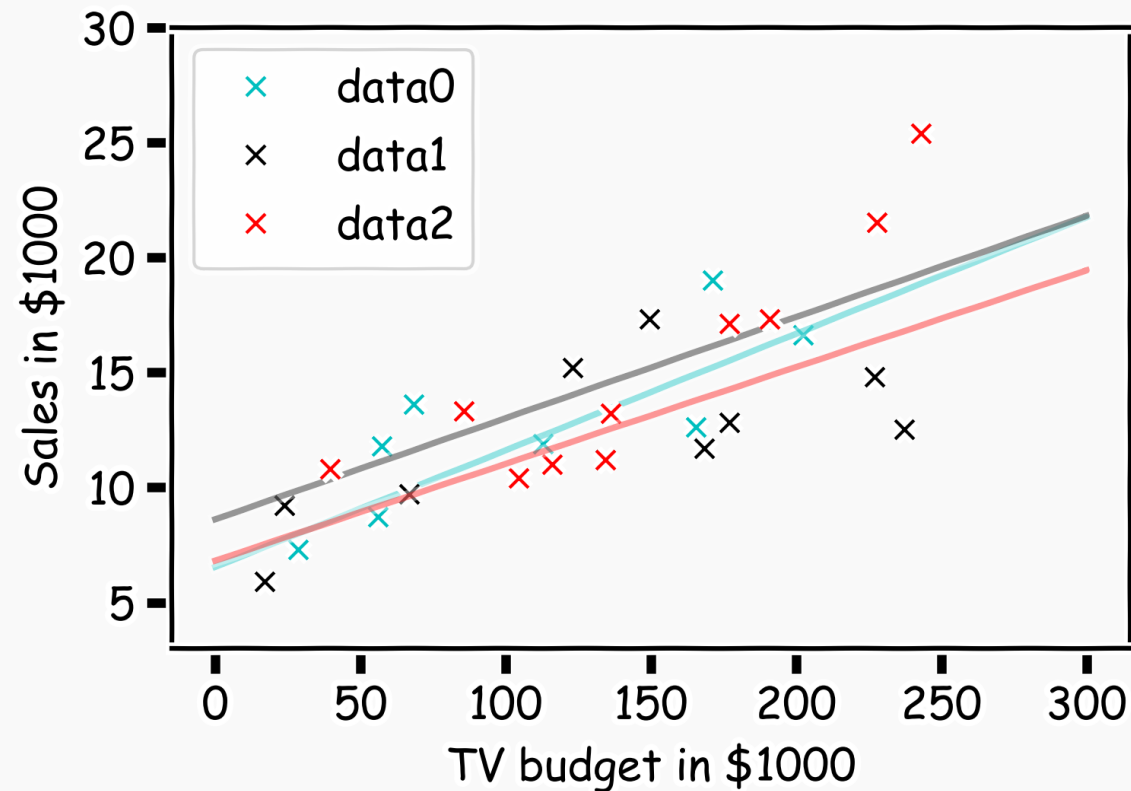
How well do we know \hat{f} ?

Here we show two different models predictions given the fitted coefficients.



How well do we know \hat{f} ?

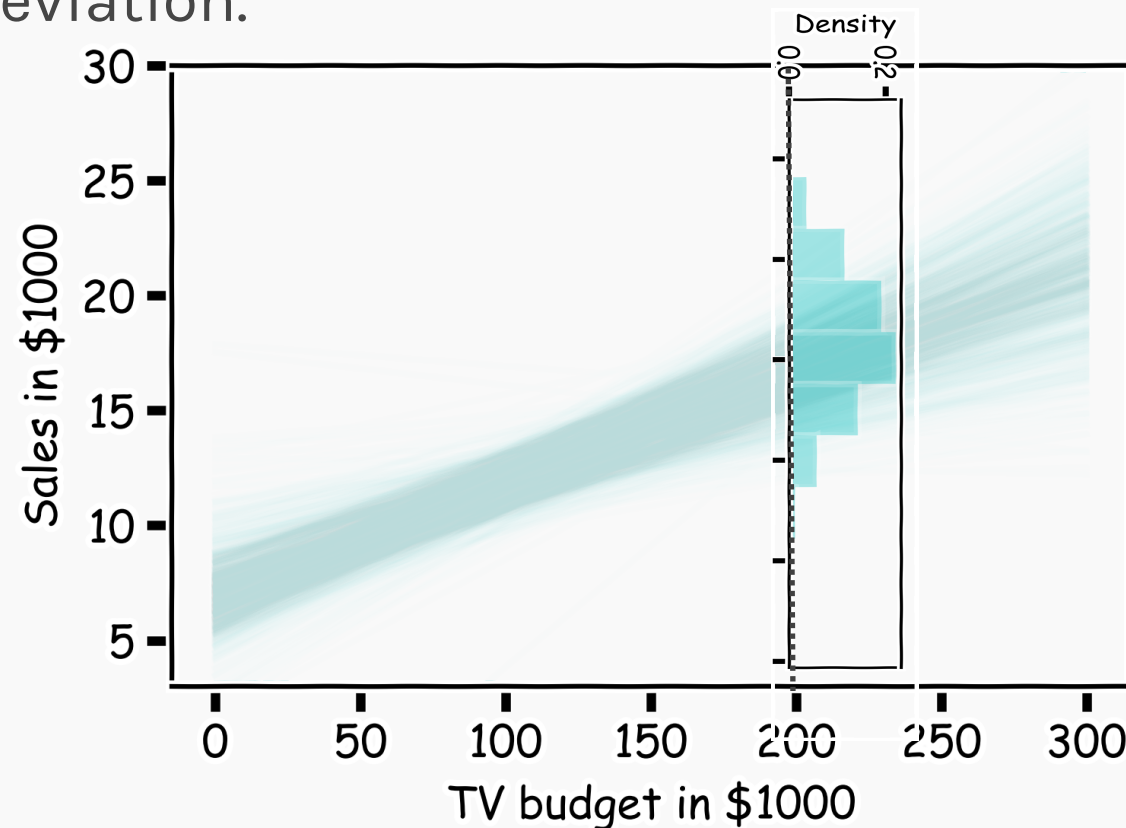
There is one such regression line for every bootstrapped sample.



How well do we know \hat{f} ?

Below we show all regression lines for a thousand of such bootstrapped samples.

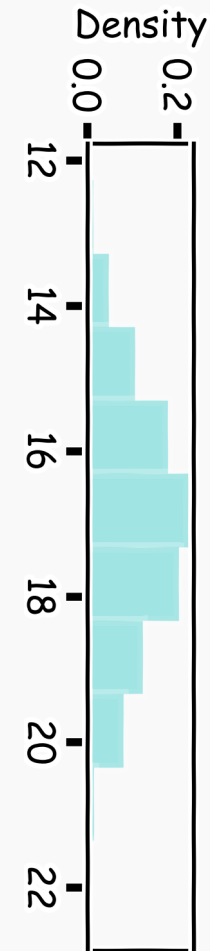
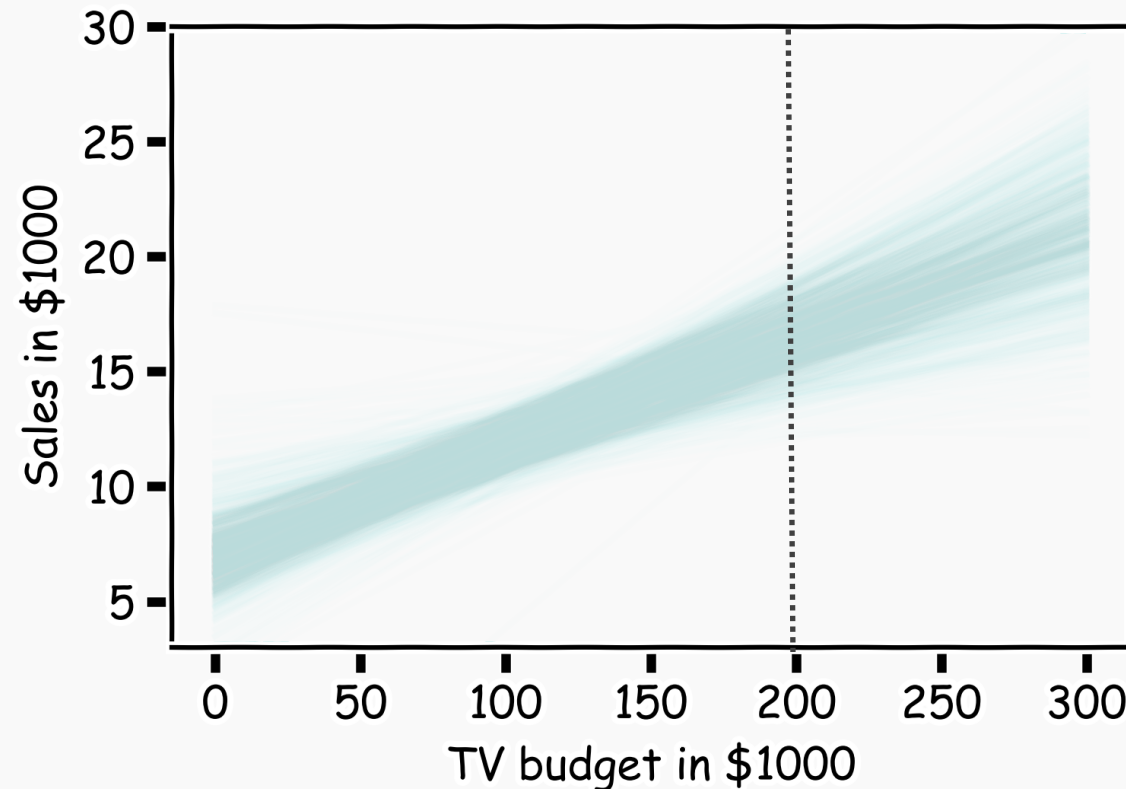
For a given x , we examine the distribution of \hat{f} , and determine the mean and standard deviation.



How well do we know \hat{f} ?

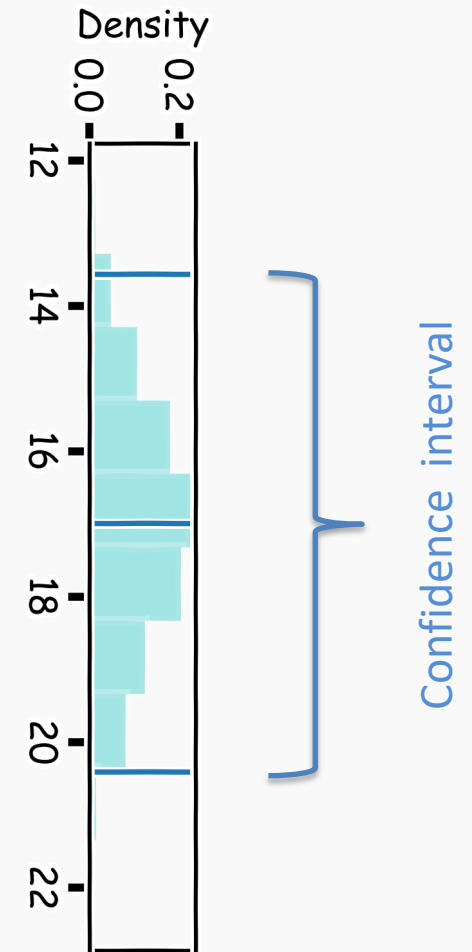
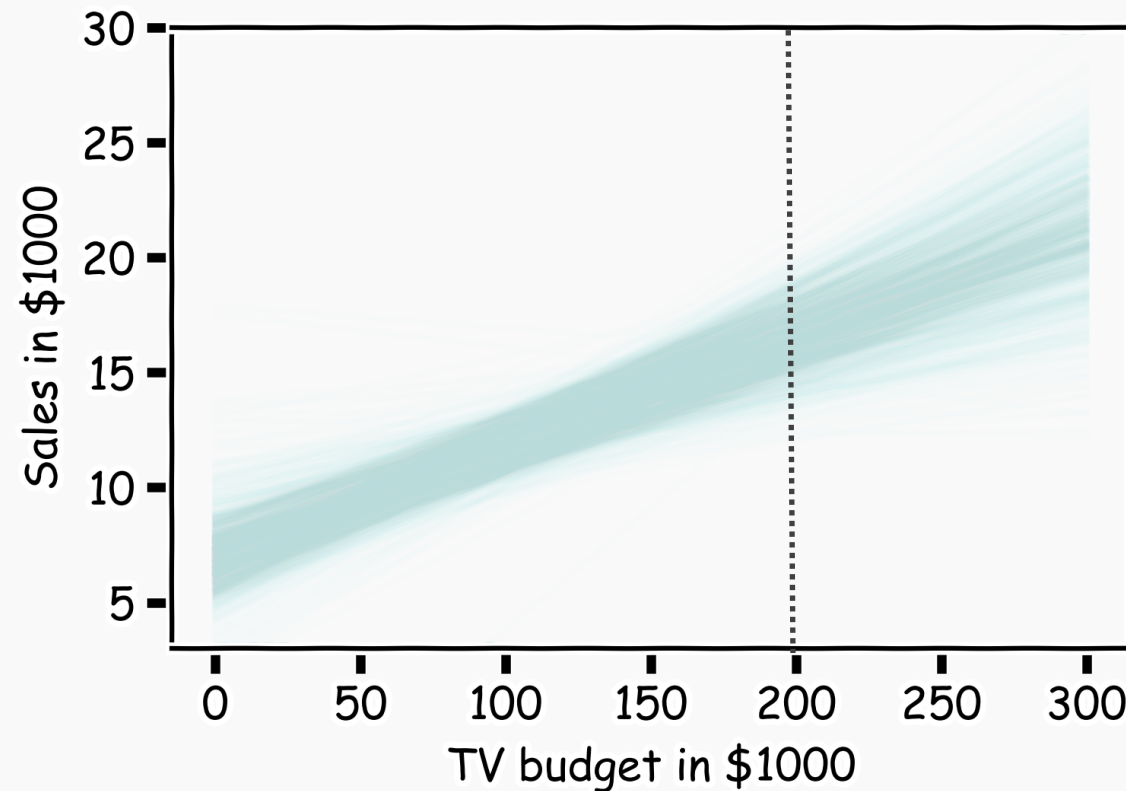
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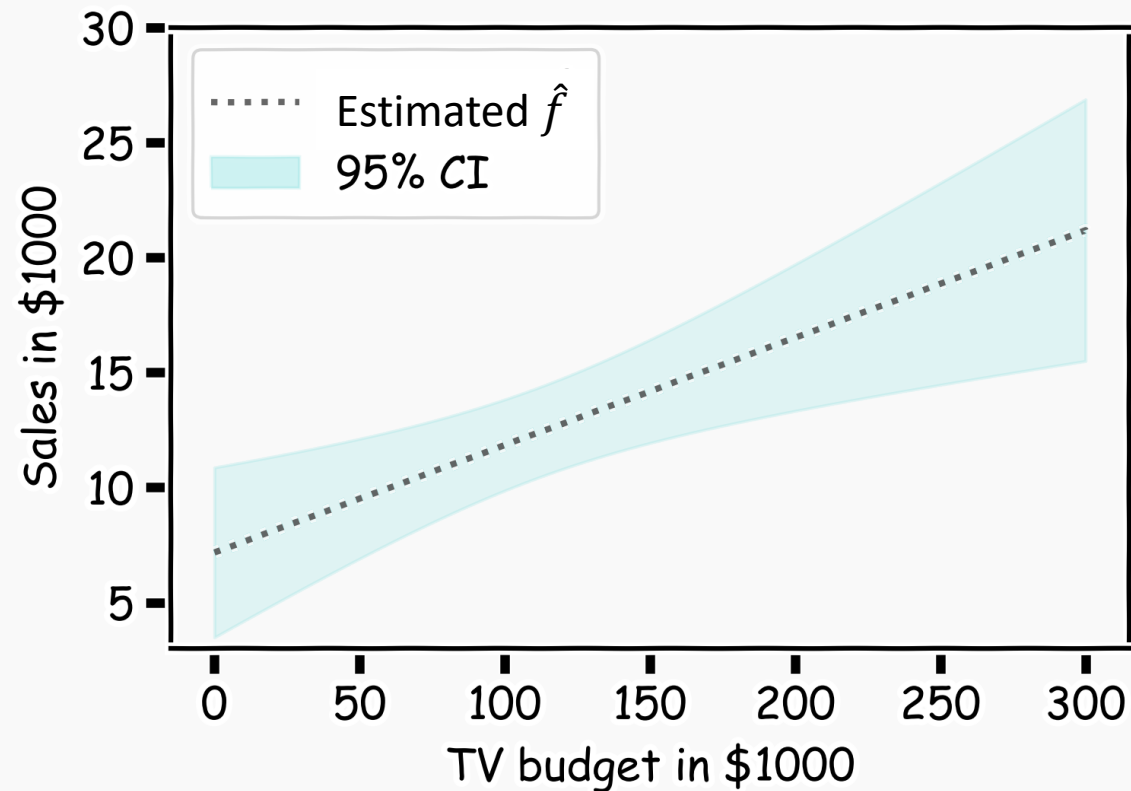
How well do we know \hat{f} ?

We determine the confidence interval of \hat{f} by selecting the region that contains 95% of the samples of $\hat{f}(x) = X \hat{\beta}$.



How well do we know \hat{f} ?

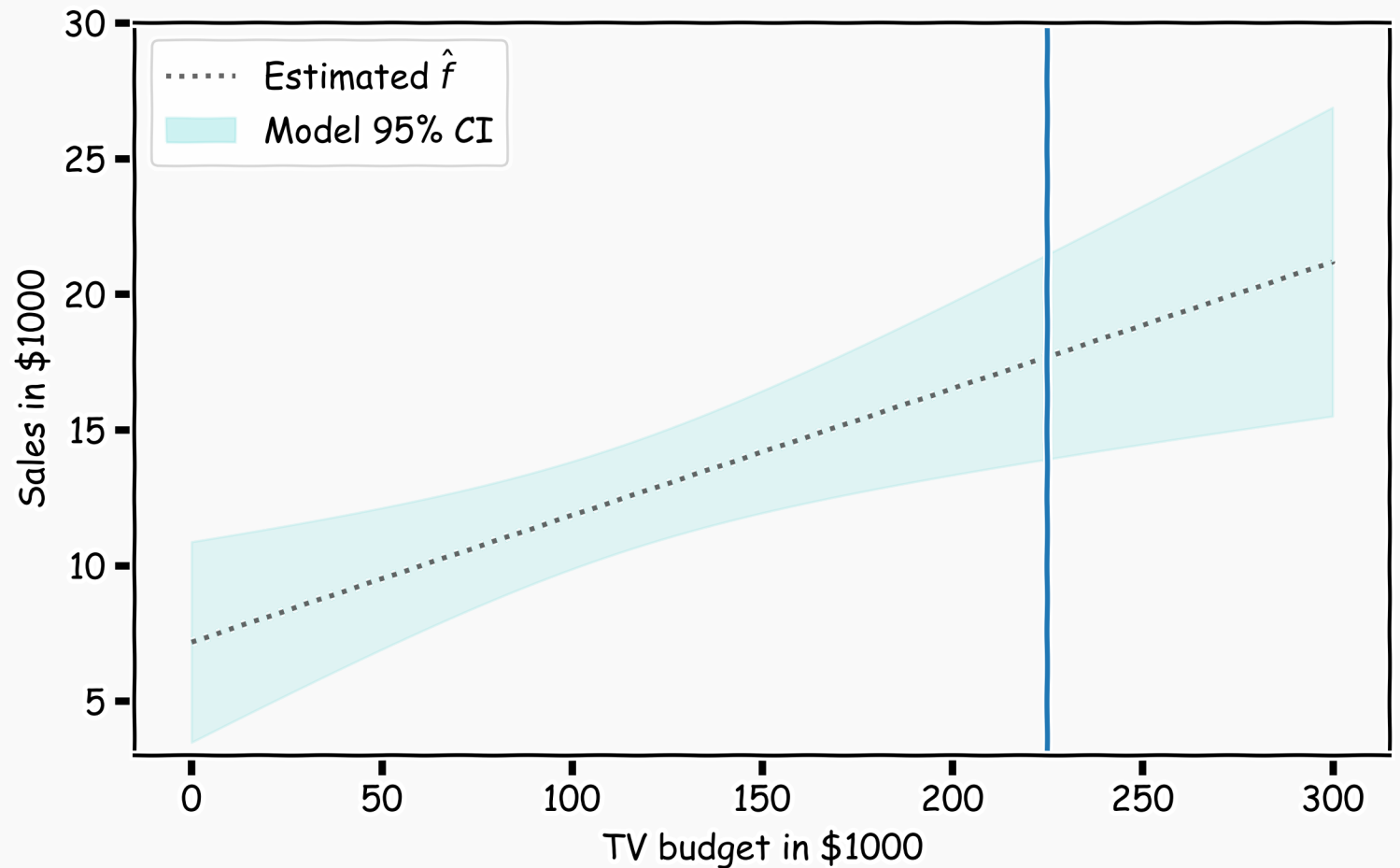
For every x , we calculate the mean of the models, $\widehat{\mu}_f$ (shown with dotted line) and the 95% CI of those models (shaded area).



Confidence in predicting \hat{y}

Even if we knew $f(x)$ —the response value cannot be predicted perfectly because of the random error in the model (irreducible error).

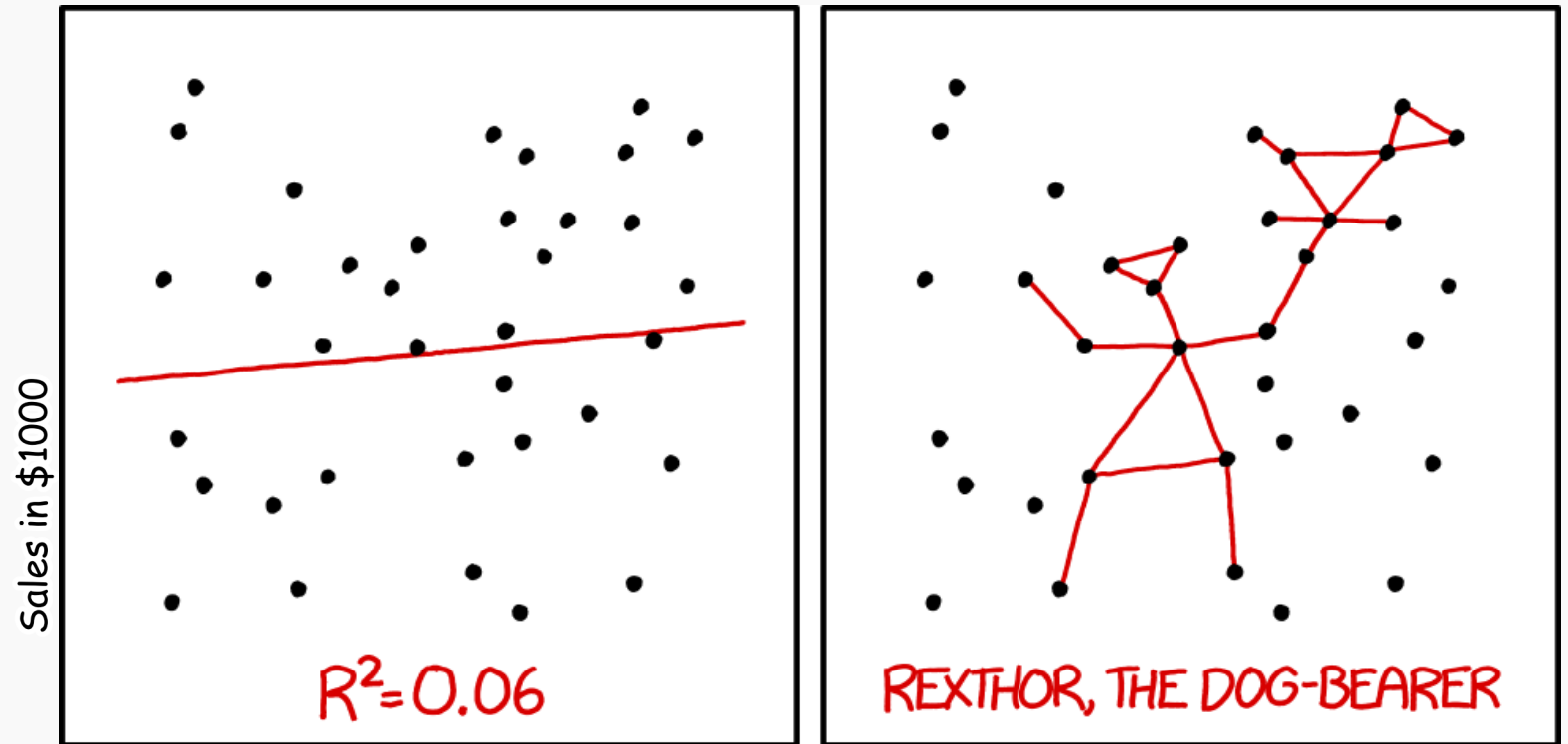
How much will Y vary from \hat{Y} ? We use **prediction intervals** to answer this question.



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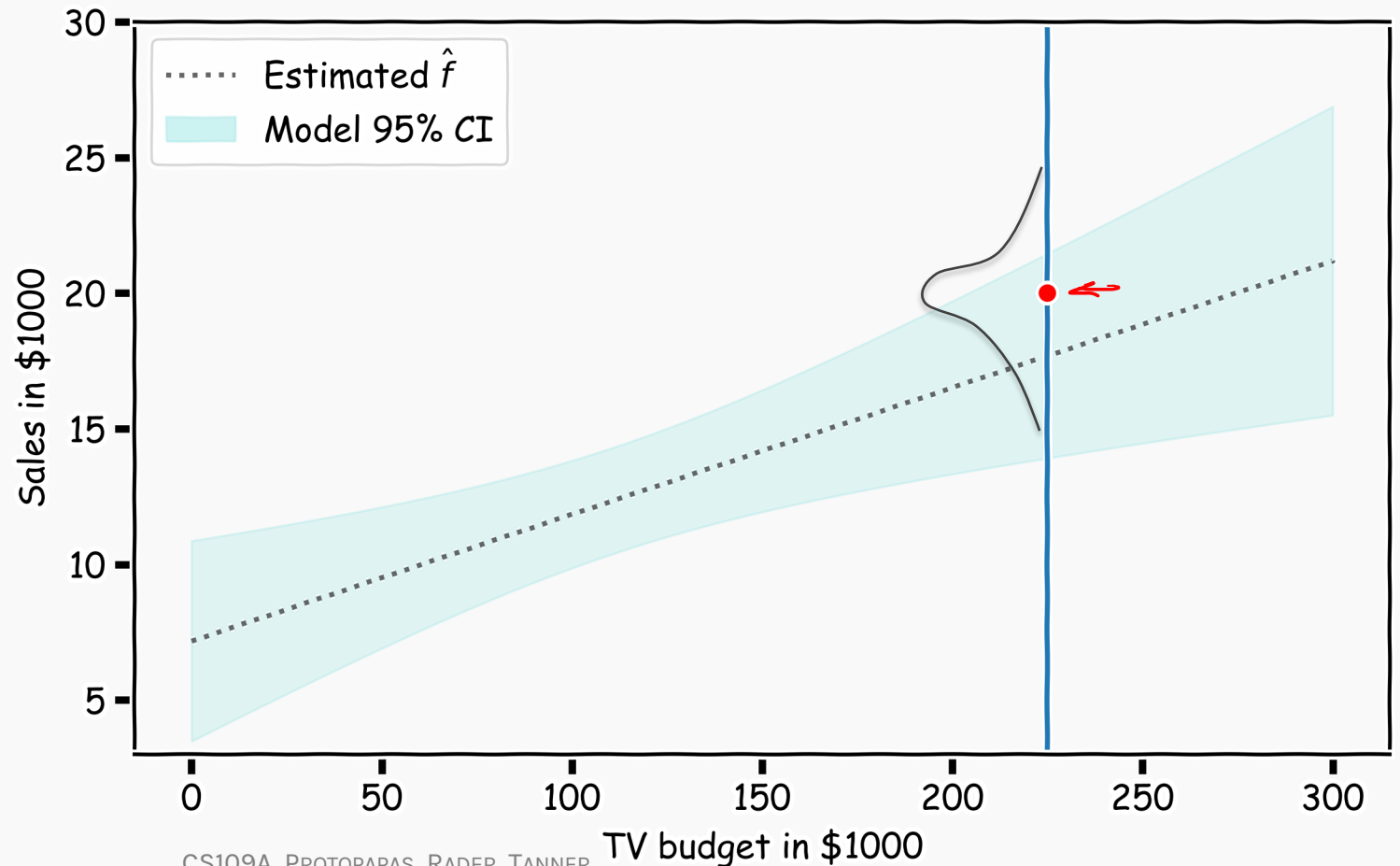
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I DON'T TRUST LINEAR REGRESSIONS WHEN IT'S HARDER TO GUESS THE DIRECTION OF THE CORRELATION FROM THE SCATTER PLOT THAN TO FIND NEW CONSTELLATIONS ON IT.

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- for a given x , we have a distribution of models $f(x)$
- for each of these $f(x)$, the prediction for $y \sim N(f(x), \sigma_\epsilon)$



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- The prediction confidence intervals are then ...

