



# Determining Income Dynamics and Optimizing Tax Rate Over Permanent Income

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April 25, 2019

(Ongoing)

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# Motivation

## Research Question

What determines income? Would it make sense to tax based on future/permanent income?

- Given the objective function, what would be the optimal tax rate?
- How would one accurately identify and predict future income?

# Decision Theory

## Payoff Function

$$\operatorname{argmax}_{D \in \mathcal{D}} \pi(Y, D, X) = B(Y(X, D), D) + C(X, D) \quad (1)$$

- Where  $B(\cdot)$  and  $C(\cdot)$  are the benefit and cost functions, and  $Y = f(X)$ .
- For a concave payoff function and differentiable  $\mathcal{D}$ , the FOC of (1) would yield:

$$B_Y(Y(X, D), D) \times Y_D(X, D) + B_D(Y(X, D), D) = C_D(X, D).$$

Choosing our payoff function as:

Permanent Income Hypothesis (Friedman)

$$c_t = \frac{r}{(1+r) + (1+r)^{-(T-t)}} \left\{ A_t + \sum_{k=0}^{T-t} \left( \frac{1}{1+r} \right)^k E_t[y_{t+k}(1-\tau_t)] \right\} \quad (2)$$

- Where  $r$  is the interest rate,  $A$  asset,  $y$  income,  $\tau$  tax rate and  $E_t[.]$  is the expected value at time  $t$ .

Why?

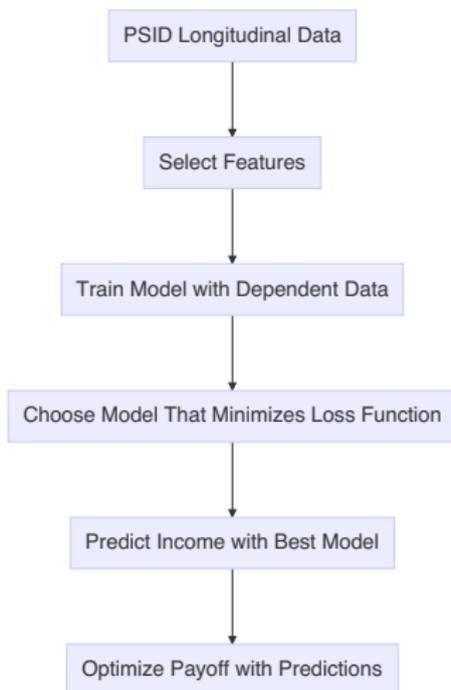
- 1 Nonlinear
- 2  $\frac{\partial^2 C}{\partial Y_d^2} < 0$ , decreasing MPC.
- 3 Defined temporally.

# Methodology

## Functional Form

$$Y = f(X) + \epsilon, \quad Y|X \sim N(0, \sigma^2) \quad (3)$$

- Typically, if  $Y$  is income, then  $\log(Y)|X \sim N(0, \sigma^2)$ .
- $Y$  is often undeterministic.
- Want to estimate  $\hat{Y} = \hat{f}(x)$ , parametrically or nonparametrically.
- Employ feature selection method (like LASSO, Random Projection, Spike-and-Slab) on sparse matrix and prediction method (like Random Forest, Boosting, SVM).

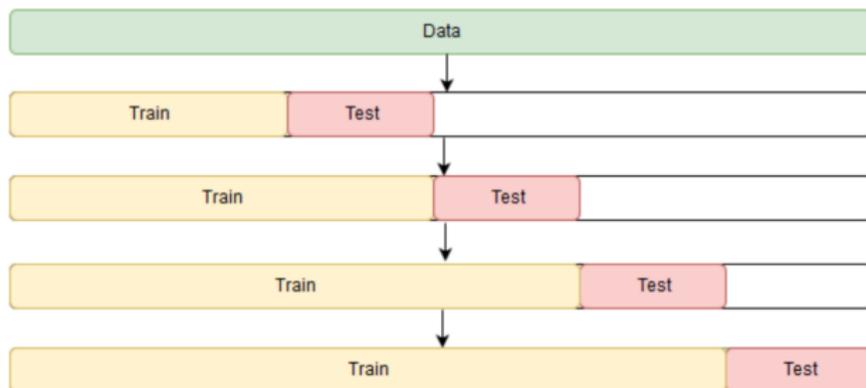


<sup>1</sup>

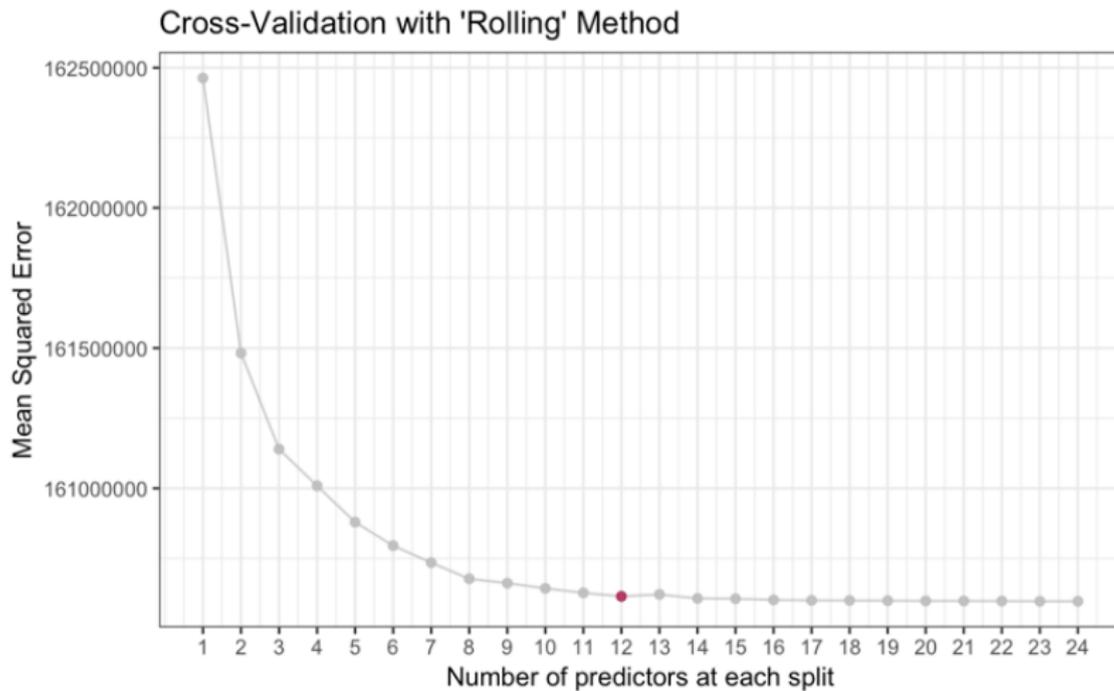
<sup>1</sup>The Panel Study of Income Dynamics (PSID) is the longest running longitudinal household survey in the world with sample of over 18, 000 individuals in 5, 000 families in the United States.

# Analysis

- Generally, we want to predict on a dataset that is independent. Or,  $Y|X \perp Y'|X'$ .
- Due to the nature of our longitudinal data, we would like train our model using a cross-validation method employed with time series data. (e.g. “Rolling” Cross-Validation).
- Roll the train and test sets forward with respect to time  $t$ .



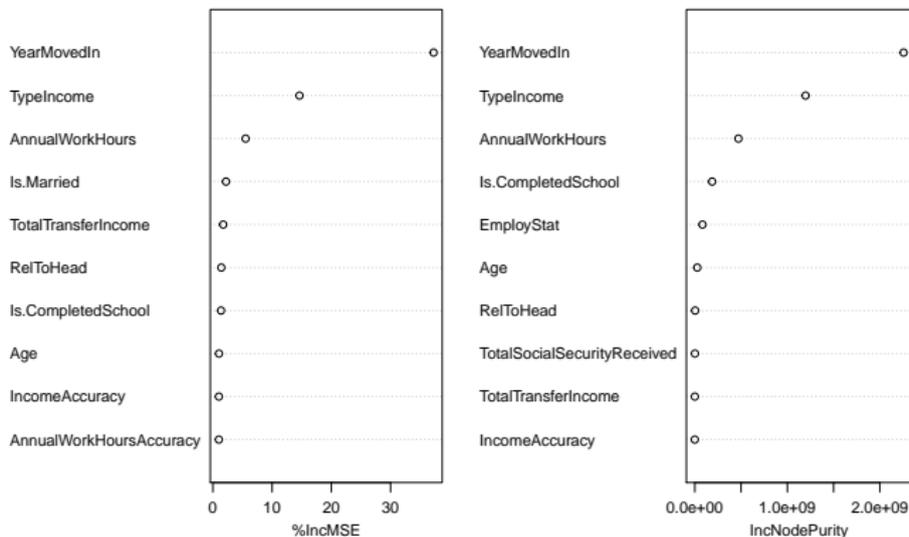
- Suppose our best model is the Random Forest:



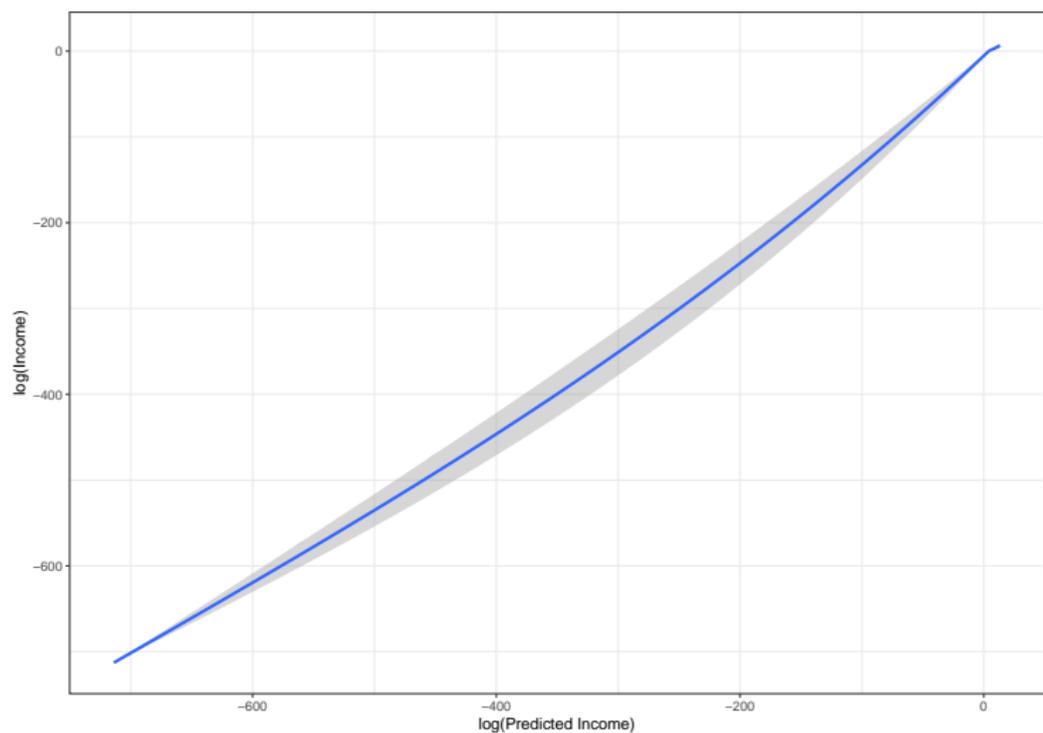
# Variable Selection

- Top 10 variables selected by Random Forest:

Variable Importance with Breiman's Method



- Comparison between actual income and predicted income.



# Optimization

- Given the predicted incomes we want to optimize the following:

$$h(\hat{y}_{t=1,\dots,T}) = \begin{cases} c_t & c_t \geq 0 \\ \tau & 0 \leq \tau \leq 1 \end{cases}$$

# Limitations

- 1 Lacking causal interpretation.
  - Honest Causal Trees (Athey, 2018).
- 2 Lacking consistency and asymptotic normality.
  - Inconsistent variance, unable to construct confidence interval. (Athey, 2018)
- 3 Hard to understand.
  - Often a black box.