Summary of Missing values handling for machine learning portfolios

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1) What are the research questions?

• How to effectively handle missing values in machine learning portfolios?

2) Why are the research questions interesting?

- Applies ML to asset pricing often encounters missing values problem.
- Drop or impute missing values are often untenable or dangerous.
- Selecting an appropriate method to handle missing values is crucial.

3) What is the paper's contribution?

- Literature on missing values in cross-sectional predictor data.
 - Prior: avoid modeling the missingness process.
 - This study: discussing the origins of missingness.
- Literature on predictor and stock return datasets.
 - Prior: 80+ predictors from the CZ open-source dataset,...
 - This study: size and breadth of datasets, diversity of statistical methods.
- Literature on using mean imputation in ML.
 - Prior: Combine large sets of predictors.
 - This study: simple imputation algorithm is helpful for transparency.

4) What hypotheses are tested in the paper?

- Hypotheses
 - Simple mean imputation for handling missing values outperforms the Expectation Maximization (EM) algorithm.
- The logic of hypotheses
 - Structure of missingness; low cross-sectional correlations; estimation noise.

5) Sample

• 159 predictors include CRSP, Compustat, and IBES analyst forecast data.

6) Regression/prediction model specification

• Baseline forecasts: principal component regressions (PCR).

7) What difficulties arise in drawing inferences from the empirical work?

• The existing results still struggle to explain the issue of dimensionality.

8) Describe at least one publishable and feasible extension of this research.

• Research on missing value handling methods based on panel data.