University of Auckland Department of Accounting & Finance

Finance 788: Research Essay

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Acknowledgements

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Abstract

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List of Equations

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1 Introduction (3 pages)

2 Literature (3 pages)

Overview of literature in asset pricing (761/751), ML application, factor pricing - very brief, 12pt, double spaced

2.1 Asset Pricing

2.2 Machine Learning

Convexity is an important concept in optimisation Monotonic ranking

2.3 Optimisation

3 Motivation

3.1 Research Question

Can neural networks, optimized to maximise financial metrics (e.g., Hedge Portfolio Excess Return, Sharpe Ratio etc.,) outperform conventional loss minimisation optimisation strategies, when predicting excess returns in individual equities and equity hedge portfolios?

3.2 Hypotheses

4 Literature Review (3 pages)

4.1 Asset Pricing

4.2 Optimisation

- Convexity
- Monotonic Ranking Function

4.3 Machine Learning

A couple of recent publications highlight the increased application of machine learning algorithms in financial contexts. Li et al., 2020 Gu et al (2020) explore the comparative use of machine learning in empirical asset pricing.

5 Methodology

The required methodology to construct the methods to build, develop, and deploy neural networks with custom objective functions.

5.1 Data

5.1.1 Global Factors Dataset

Hou et al., (2020) use an extensive data library to assess 452 anomalies across anomalies literature. Their analysis informs which abnormalities drive the cross section of expected returns. Most abnormalities fail under current standards of empirical finance when using a single hurdle test of absolute t-stat greater or equal to 1.96. Firstly, the paper finds economic fundamentals take precedence over trading frictions in explanatory power, statistical and economic significance. Secondly, micro-caps account for anomalies disproportionately, leading to NYSE breakpoints, value-weighted returns in both portfolio sorts and cross-sectional regressions with weighted least squares. Lastly, arguments in improving anomalies literature credibility follow a closer alignment to economic theory as the field persists to be statistical in nature. Overall, capital market efficiency is higher than expected. Jensen et al., 2021 use the above dataset to explore hierarchical bayesian models of alphas emphasising the joint behaviours of factors, and provide an alternative multiple testing adjustment, more powerful than common methods. Jensen et al., adapt the global dataset to focus only on one-month holding periods for all factors, only include most recent accounting data (quarterly or annually) and add 15 new factors. Section 9.7 describes factor composition, resources for data acquisition, and summary statistics. The complete global dataset has 406 characteristics, a superset of the original 153 in Jensen et al., with 2,739,928 firm-year observations, from January 1st 1961 to December 31st 2020. Subsequently, the complete dataset has 1.112 billion data points. One month lead excess returns is the designated target variable for prediction as will inform the construction of hedge portfolios to assess relative performance between optimisation functionalities. The exhaustive nature and accessibility of the global dataset

makes it well-suited for exploring optimisation functions, maximising renown financial metrics, in deep neural-networks.

5.1.2 Processing

Neural networks demand the partitioning of the dataset into training, validation, and testing subsets. The initial training, testing, and validation sets consist of 1031516, 706908, and 1001504 global equity firm-year observations across 406 features, respectively. The division of subsets is chronological with firm-year observations [1961-1990), [1990-2000), [2000-2020] for training, validation, and testing, respectively. Two reasons rationalise the reduction in the number of factors from 406 in the Jensen et al., (2021) superset to 153, and the removal of firm-year observations with Micro or Nano size grouping ¹ designations. Firstly, the retention of equities between 20th to 100th percentiles appeals to economic significance. The composition of aggregate market capitalisation is mostly from their contribution. Additionally, their higher liquidity increases the likelihood of analyst coverage and portfolio inclusion. Secondly, factor reduction appeals to parsimony in aligning explanatory variables to prior studies. The training, testing, and validation sets consist of 532218, 294581, and 531461 global equity firm-year observations across 160 features after above revisions, respectively. Subsequently, the revised dataset has 217,321,600 data points. Tables 5, 7, and 6 in section 9.7.3 describe summary statistics and factor retention after subset revision.

5.1.3 Cloud Infrastructure

Cloud-centric computational products execute data processing and analysis. Google Cloud Platform Cloud Storage buckets and Compute Engine virtual machine (VM) instances manage large datasets and build, train, and evaluate deep neural networks, respectively. Cryptographic network protocols, mostly secure shells, establish remote connectivity between local and remote infrastructure to communicate and execute commands.

¹Mega, Large, and Small remain, reflecting equities with market capitalisations greater than the 80th, 50th, and 20th percentile of all NYSE stocks, respectively. Micro equities reside between the 1st and 20th percentiles, with Nano between below the 1st percentile. The percentiles are non-overlapping and value weighted by market capitalisation on the New York Stock Exchange (NYSE).

However, use of cloud computing does not resolve all resource constraints. ² Section 9.5 elaborates on cloud centric computational approaches and further technologies.

5.2 Neural Networks

5.2.1 Normalisation & Feature Encoding

Deep neural networks require tensors as inputs for fitting, training, and evaluating data. A tensor is a mathematical object describing the physical properties of an object with multilinear relationships between sets of algebraic objects related in vector space. Furthermore, transformation laws govern tensors. Therefore, a tensor is considered an n dimensional array in conjunction with associated transformation laws. The dataset, also known as the feature matrix (\mathbf{X}) , must take the form of a tensor. The enumeration of a tensor conversion process for each data subset follows:

- 1. Identify target variable(s) for fitting, validation, and prediction.
- 2. Configure dataset into a series of tensor slices
- 3. Shuffle instances in convert dataset to promotes better training as accommodates randomness
- 4. Instantiate an input tensor layer and use to extract required normalisation or encoding layer per feature
- 5. Normalise each numerical feature to zero mean/unit variance, and encode ³ each categorical feature, using encoding layers to form encoded features
- 6. Combine instantiated input tensors (4) into a set of inputs tensors, serving as the inputs fro a configured neural network
- 7. Concatenate all encoded features into an aggregate input tensor for an unconfigured neural network. The concatenation serves as the input layer when configuraing a neural network.

The revised datasets consist of eight categorical features ⁴ and 152 numerical features.

²Resource constraints inhibit exploration of the entire dataset within reasonable timeframes at reasonable costs. The most material inhibitions are the inability to explore all 406 factors for all size groupings, the reduction in level of precision for numerical features, and the ability to shuffle training sets at lengths greater than or equal to the input sets when training neural networks. However, the reasons necessitating factor reduction and size grouping exclusion stated in section 9.7.2 superseed resource constraints.

³Encoding transforms categorical features instances into a series of binary variables. These may be one-hot encoded, and/or stored in sparse tensors, depending on input and desired application.

⁴Descriptions of the eight categorical features follow. **size_group**, the aforementioned size grouping in section . **permno**, permanent unique firm identifier from CRSP. **permco**, permanent unique issue identifier from CRSP. **crsp_shrcd**, CRSP share code. **crsp_exchcd**, Compustat stock exchange code. **sic**, Firm SIC industry. **ff49**, Classification of stocks in 49 industry groups, based on SIC codes and the methodology in F. Fama & K. French 1997, with the addition of the software industry. **adjfct**, Share Adjustment Factor

5.3 Neural Network Architecture

5.3.1 Activation Functions, Linear Threshold Units, & Perceptrons

Artificial Neural Nets (ANN) frequently outperform other machine learning algorithms on large and complex problems. Linear threshold units (LTU) compose neural networks, feeding the weighted sum of input values (1) into an activation (step) function (2). A perceptron is a single layer of LTUs connected to every input, suitable for both regression and classification tasks. Perceptrons utilize a training algorithm to assess the strength of connections between perceptrons in consideration of errors. A perceptron is fed one training instance sequentially, making predictions for each instance. For every output LTU that produced a wrong prediction, it re-enforces the connection weights using the perception learning rule (3) from the inputs that would have contributed to the right prediction. One input perceptron, multiple hidden perceptrons, and an output perceptron create a Multi Layer Perceptron (MLP). A non-linear activation function ⁵ (i.e., Logistic (4), Rectified Linear Unit (ReLU) (5)) replaces the step functions for an LTU, in each perceptron, in an MLP. A shared activation function replaces the individual activation functions in the output layer to enable exclusive classification or regression.

$$\mathbf{z} = \mathbf{w}^T \cdot \mathbf{x}$$
 (1) $h_w(\mathbf{x}) = step(\mathbf{z})$ (2) $w_{i,j}^{\text{next step}} = w_{i,j} + \eta(\hat{y}_j - y_j)x_i$ (3)

$$\sigma(\mathbf{z}) = \frac{1}{1 + exp(-\mathbf{z})} \quad (4) \quad ReLU(\mathbf{z}) = max(0, \mathbf{z}) \quad (5) \quad Linear(\mathbf{z}) = \mathbf{z} \quad (6)$$

Where

- $w_{i,j}$ is the connection weights between the ith input neuron and the jth output neuron.
- x_i is the ith input value of the current training instance.
- \hat{y}_j is the output of the jth output neuron for the current training instance.
- ullet y_j is the output of the jth output neuron for the current training instance.
- η is the rate.

⁵There are several activation functions, each with different strengths and weaknesses

5.3.2 Model Configuration

Figure 1 visualises a standard neural network topography

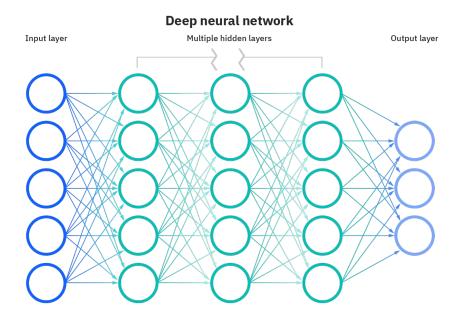


Figure 1: Standard Neural Network Topography (Source: IBM)

The dots and lines represent nodes and connections between nodes, respectively. The architecture of the network is derivative of intended use. The set of The linear activation function (6) is most suitable output layer activation function for regression task as predict values between $(-\infty, \infty)$ directly, suitable for evaluating excess returns. Figure 2 illustrates the configuration for this project (work in prgoress).

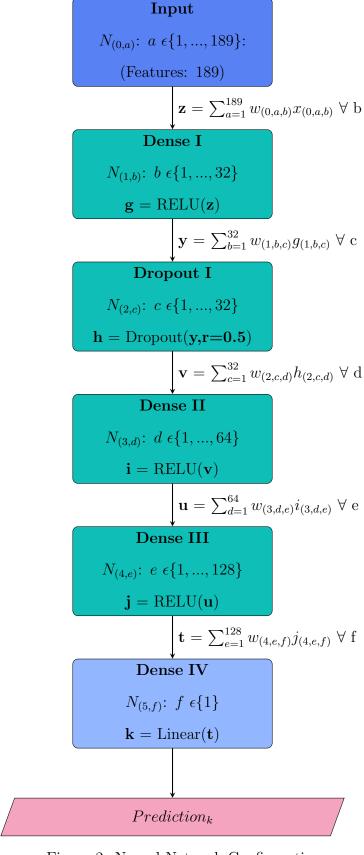


Figure 2: Neural Network Configuration

10 epochs, 207 runs per epochs.

5.4 Brief Description of Remaining Methods

- Use stochastic gradient descent, train model with 10 epochs (2709 runs per epoch) for each of the six loss functions (tensorflow mse, custom mse, sharpe ratio, sharpe ratio mse, hedge portfolio mse, information ratio). Validate with validation set and evaluate with testign set.
- Make predictions using entire testing set for each model. Merge predictions with fama french factors on month
- Perform newy west regressing hedge portfolio returns on fama factors. Predictions grouped by month. hedge formed by mean of top decile mean of bottom decile creating hedge return for the month.
- Second set of regressions does a panel regression accounting Heteroskedasticity Robust Covariance as Fama MacBeth did not execute due to rank deficiencies.

5.5 Loss Functions & Performance Metrics

5.5.1 Modelling, Loss, and Optimisation

We summarize the theory surrounding predictive modelling, loss functions, and optimisation algorithms. These functions train models by comparing predictions to realized observations using optimisation algorithms to minimize the loss function. We examine a linear model as our predictive model (7). Mean square error (8) and Gradient Descent (GD) are basic examples of a loss function and optimisation algorithm, respectively.

$$\hat{y} = mx_i + b \tag{7}$$

$$f(y, (mx_i + b)) = \frac{1}{n} \sum_{i=1}^{n} (y_i - (mx_i + b))^2$$
(8)

Firstly, gradient descent takes the partial derivatives of the loss function, with the respect to the parameters in our predictive model. In our example, equations 9 and 10 are the partial derivatives for the mean square error loss function.

$$\frac{\partial f(y, (mx_i + b))}{\partial m} = \frac{1}{n} \sum_{i=1}^{n} -2x_i (y_i - (mx_i + b))^2$$
(9)

$$\frac{\partial f(y, (mx_i + b))}{\partial b} = \frac{1}{n} \sum_{i=1}^{n} -2(y_i - (mx_i + b))^2$$
 (10)

Secondly, the algorithm explores epochs, using a learning rate to update parameters to move in the opposite directions of the partial derivatives until settling in a local minima. This extrema is the optimisation of the loss function, quantifying the accuracy of the predicative model. Ordinary Least Squares (OLS) regressions is an extension of the linear model prevalent in asset pricing. The mathematical rigor and suitability of the OLS estimator section 9.2. The theory underlying OLS informs the common practice in minimising of the sum of least squares when evaluating prediction performance. The mathematical tractability, in accordance with the aforementioned assumption, frame our thinking surrounding the derivation of custom loss functions.

5.5.2 Automatic Differentiation

5.5.3 Loss & Metric Definition

5.6 Hedge Portfolio Construction

5.7 Regression Analysis

5.8 Loss Functions & Performance Metrics

Table 1 emphasises the separation between training and validation datasets.

Variable	Description	$MSE(y, \hat{y})$	$HP(y,\hat{y})$
θ	Estimation Training	$\hat{ heta}_{MSE}$	$\hat{ heta}_{HP}$
λ	Validation	$\hat{\lambda}_{MSE}$	$\hat{\lambda}_{HP}$

Table 1: Objective (MSE: Mean Square Error, HP: Hedge Portfolio)

5.8.1 Mean Square Error (MSE)

Section 5.5.1 outlines advantages to Ordinary Least Squares. Subsequently, MSE serves as a baseline for loss function and performance metric comparisons. The following function (11) and partial derivative (12) describe Tensorflows's Mean Square Error implementation, both from in-built and custom contexts. Python classes describe equation 11 to enable Tensorflow's automatic differentiation capabilities, approximating the partial derivatives of the loss function (12) with numerical methods. Please note the use of Hadamard exponentiation $(x^{\circ n})$ as an element-wise operation.

$$f(y, X^T \hat{\theta}) = \frac{\vec{1}}{\vec{1}^T \vec{1}} (y - X^T \hat{\theta})^{\circ 2}$$
(11)

$$\frac{\partial f(y, X^T \hat{\theta})}{\partial \hat{\theta}} = \frac{\vec{1}}{\vec{1}^T \vec{1}} (-2(y - XT \hat{\theta})^{\circ 1})$$
(12)

5.8.2 Hedge Portfolio

Hedge portfolios rely on monotonic ranking functions for optimisation as their monotonic nature preserves or reverses a given ordered set. The analysis cross-section of one-month lead portfolio excess returns using monotonic functions

$$R(y_{i,t}) \tag{13}$$

The ranking function $(R(y_{i,t}))$ and thresholds (u,v) form subsets of long and short portfolios. Long (L) or Short (S) sets include excess returns conditioned on the associated monotonic ranking given a threshold, bound by the cardinality of the excess return vector (|y|). The subsequent truth sets mathematically express aforementioned time-series hedge portfolios.

$$L = \{y_{i,t} | R(y_{i,t}) \le u\}$$

$$S = \{y_{i,t} | R(y_{i,t}) \ge v\}$$

$$0 < u \le |y|$$

$$0 < v \le |y|$$

$$u < v$$

Equation 14 describes hedge portfolio lead excess returns (H_t) at a given time (t).

$$H_t = \frac{1}{|L|} \sum_{i \in L} y_{i,t} - \frac{1}{|S|} \sum_{i \in S} y_{i,t}$$
 (14)

Figure 3) illustrates an approximate linear monotonic ranking function with a sample of 100 uniformly distributed excess returns between -10% and 10%. Boundary conditions u and v are set to 20 and 80, respectively. Subsequently, excess returns above (below) the green (blue) dotted line belong to the long (L) (short (S)) set.

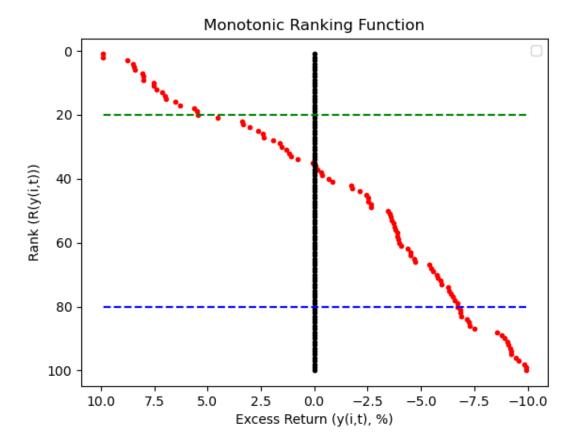


Figure 3: Approximate Linear Monotonic Ranking Function

The permutations in monotonic ranking functions, and subsequent hedge portfolios, are endless. This research essay develops a monotonic ranking function proportionally weighting one month lead excess returns (15). Therefore, equation 16 defines the loss function.

$$R(\hat{y}) = W$$

$$W := \frac{\hat{y}}{\mathbf{1}\hat{y}}$$
(15)

$$\hat{y} = X^T \hat{\theta}$$

$$f_{\hat{\theta}}(X) = (\frac{X^T \hat{\theta}}{\vec{1} X^T \hat{\theta}})^\top X^T \hat{\theta}$$
(16)

The above loss function is differentiable using symbolic mathematic as shown in equation

17.

$$\frac{\partial f_{\hat{\theta}}(X)}{\partial \hat{\theta}} = \frac{\partial ((\frac{X^{T}\hat{\theta}}{\vec{1}X^{T}\hat{\theta}})^{\top}X^{T}\hat{\theta})}{\partial \hat{\theta}}
\frac{\partial (f_{\hat{\theta}}(X))}{\partial \hat{\theta}} = \frac{1}{(\hat{\theta}^{\top}X\vec{1})}XX^{\top}\hat{\theta} + \frac{1}{\vec{1}X^{\top}\hat{\theta}}XX^{\top}\hat{\theta} - \frac{1}{(\hat{\theta}^{\top}X\vec{1})^{2}}\hat{\theta}^{\top}XX^{\top}\hat{\theta}X\vec{1}$$
(17)

Our research Subsection 5.5.1 explains the theory supporting loss minimisation. Applying gradient descent methods to the product of the loss function and scaler of -1 transforms the minimisation to maximisation. This transformation leads to finding the argmax of maximisation function with respect to $\hat{\theta}$ (18). The aforementioned transformation is simply and suitable for exploration in the context of the research intent. More sophisticated methods exist for maximisation such as reinforcement learning (??).

$$\underset{\hat{\theta}}{\operatorname{argmax}} : (\frac{X^T \hat{\theta}}{\vec{\mathbf{1}} X^T \hat{\theta}})^\top X^T \hat{\theta}$$
 (18)

5.9 Conventional Asset Pricing Models

Conventional asset pricing methodologies persist in academic literature. The use of the Capital Asset Pricing Model (CAPM) persists, regardless of the identifiable shortcomings in market proxies and empirical failings invalidating use (Fama and French, 2004). Nonetheless, this research essay uses the model as a performance metric for comparative purposes.

$$R_{i,t} - R_{f,t} = \alpha_{i,t} + \beta_{i,t}^1 (R_{M,t} - R_{f,t})$$
(19)

E. Fama and K. French (1992) validate the explanatory power of size and value (book-to-market) factors in their ability to capture the cross-sectional variation in average stock returns, in association with market risk, size, leverage, book-to-market, and earnings-price ratios. E. Fama and K. French further their analysis on the common characteristics between stocks and bonds (Fama and French, 2021)⁶, and add two additional factors to

⁶Reprinted. Originally published in 1993

consider profitability and investment. The main combinations are the Fama French Three (FF3) (20) and Five (FF5) (21) models. E. Fama and K. French consider a momentum factor on international stock returns in subsequent years (Fama and French, 2012). The omission of momentum from the models stand.

$$R_{i,t} - R_{f,t} = \alpha_{i,t} + \beta_{i,t} (R_{M,t} - R_{f,t})$$
(20)

$$R_{i,t} - R_{f,t} = \alpha_{i,t} + \beta_{i,t} (R_{M,t} - R_{f,t})$$
(21)

where

- $R_{i,t} R_{f,t}$: Portfolio Excess Return on the market for a given portfolio and time, value-weighted using all incorporated US CRSP firms incorporated in the US, listed on the NYSE, AMEX, or NASDAQ.
- $\alpha_{i,t}$: Jensen's alpha indicating mispricing in the asset.
- β_t^1 : Market Risk Factor (co-efficient)
- β_t^1 : Size Factor (co-efficient)
- β_t^2 : Value Factor (co-efficient)
- β_t^2 : Value Factor (co-efficient)
- β_t^3 : Profitability Factor (co-efficient)
- $(R_{M,t} R_{f,t})$: Market Risk Premium
- SMB_t : Size Premium (small minus big) is the difference in average return between nine small stock and nine large value-weighted portfolios.
- HML_t : Value Premium (high minus low) is the difference in average return between two value and two growth value-weighted portfolios.

- RMW_t : Profitability Premium (robust minus weak) is the difference in average return between two robust operating profitability and two weak operating profitability value-weighted portfolios.
- CMA_t : Investment Premium minus aggressive is the difference in average return on the two conservative and two aggressive investment portfolios

K. French continues to maintain FF3 and FF5 related datasets (French, 2021) E. Fama, with J. MacBeth, developed the Fama-MacBeth regression (Fama and MacBeth, 1973) to estimate factor loadings and prices. The methodology is a two-stage estimation process, similar for estimating factor loadings, and prices, for a given portfolio. The first step requires determining each asset's β exposures by regressing each of n asset returns against m proposed 22. The second step determines the risk premium (factor pricing) for each asset by regressing all asset returns for each of T periods against previously estimated β s (23).

$$R_{n,t} = \alpha_n + \sum_{f=1}^F \beta_{n,F_f} F_{f,t} + \epsilon_{n,t}$$
(22)

$$\forall n \in \{1, ..., N\}$$

$$R_{i,t} = \gamma_{t,0} + \sum_{f=1}^{F} \hat{\gamma_{t,f}} \hat{\beta_{i,F_f}} + \epsilon_{i,t}$$
(23)

$$t\epsilon\{1,...,N\}$$

Where

- $R_{n,t}$: Return for an asset (n) at a time (t).
- α_n : Jensen's alpha for an asset (n) implying mispricing.
- β_{n,F_f} : An asset's (n) exposure to a factor (f)
- \bullet $\hat{\beta_{i,F_f}}$: Estimated factor loading for a factor (f) from regression of asset (i)

- $F_{f,t}$: Risk factor (f) at a given time (t) e.g., SMB, HML etc.,
- $\epsilon_{n,t}$: Residual for an asset at a time (t)
- $\gamma_{t,f}$: Factor pricing for a factor (f)

The residuals of risk-return regressions and co-efficient 'fair game' observations are consistent with the efficient market hypothesis ((Samuelson, 2016), (Fama, 1995), (Fama, 1963), (Fama, 1965) (Fama, 2021)). Subsequently, this research essay uses the excess one month lead returns of the hedge portfolio (??)), Capital Asset Pricing Model (19), Fama-French Three Factor Model (20), and Fama-French Five Factor Model (21) as performance metrics for each loss function. Table 2 outlines the proposed]performance metrics and loss functions.

Performance Metric	Loss Function	
Hedge Portfolio Mean	Hedge Portfolio Excess Returns	
Hedge Portfolio FF3	Sharpe Ratio	
Hedge Portfolio FF5	Information Ratio	
Sharpe Ratio	Validation	
Information Ratio	Validation	
Mean Square Error	Mean Square Error	

Table 2: Proposed Loss Functions and Performance Metrics

Hou et al (2021) provide the required data to derive these performance metric and loss functions.

5.9.1 Sharpe Ratio

Nobel Laurette William F. Sharpe (1994) introduced the Sharpe Ratio (24) as a measure for risk-adjusted returns where $\mathbb{E}[R_a - R_f]$ is the expectation for excess returns and $\sigma(R_a)$

is the standard deviation of excess returns.

$$SR = \mathbb{E} \frac{[R_a - R_f]}{\sigma(R_a)}$$

$$\sigma = \sqrt{\sum_{i=1}^n \frac{(R_a) - \bar{R}_a)^2}{n}}$$
(24)

The matrix notation for the Sharpe Ratio loss function () follows:

$$SR = 1\partial SR \tag{25}$$

5.9.2 Information Ratio

The Information Ratio (IR) is another risk-return measure (26), evaluating the performance of a portfolio compared to a selected benchmark e.g., an exchange traded fund. $\mathbb{E}[R_a - R_b]$ is the excess return on the benchmark. $\sigma(R_a - R_b)$ is the standard deviation of the difference between the portfolio and benchmark known as the tracking error.

$$IR = \mathbb{E}\frac{[R_a - R_b]}{\sigma(R_a - R_b)} \tag{26}$$

Both Sharpe and Information ratios supplement hedge portfolio-related loss functions and performance metrics.

6 Results

6.1 Performance Matrix

Loss Function	Mean	Sharpe Ratio	Treynor
mean_squared_error	-0.018481	-0.725234	0.193231
$custom_mse$	-0.023323	-1.062051	0.391186
custom_hp	0.011715	1.292756	1.962854

Hlavac, 2018 enables the tables in the regression analysis.

6.2 Model Accuracy

6.2.1 Model Loss

Table X illustrates the performance of each model during training, validation, and testing.

Table 3 informs the accruracy of predictions

	Dependent Variable:One Month lead Excess Return		
	In-Built MSE	Custom MSE	Custom HP
predict	0.033***	0.045***	0.069***
	(0.002)	(0.002)	(0.005)
Observations	531,461	531,461	531,461
R^2	0.002	0.003	0.001
Adjusted \mathbb{R}^2	0.002	0.003	0.001
Residual Std. Error	0.137	0.137	0.137
	$(\mathrm{df} = 531460)$	$(\mathrm{df} = 531460)$	$(\mathrm{df} = 531460)$
F Statistic	486.597***	461.414***	161.031***
	(df = 1.0; 531460.0)	(df = 1.0; 531460.0)	(df = 1.0; 531460.0)

Note:

*p<0.1; **p<0.05; ***p<0.01

Table 3: Explanatory Power of Neural Networks Predictions in explaining One Month Lead Excess Returns and validate model predictive capabilities

Fama MacBeth Regressions using individual securities. PooledOLS replaces Fama MacBeth in this instance.

- 6.3 Hedge Portfolio Performance
- 6.3.1 Inbuilt Mean Square Error
- 6.3.2 Custom Mean Square Error
- 6.3.3 Custom Hedge Portfolio

7 Discussion

8 Conclusion

8.1 Contributions

8.2 Further Research

8.3 Limitations

Further research will rely on addressing severals limitation.

8.3.1 Resources

The lack of resources to explore the entire dataset within reasonable timeframes, and at reasonable costs. The most material inhibitions are the inability to explore all 400 factors for all size groupings, the reduction in level of precision for numerical features, and the ability to shuffle training sets at lengths greater than or equal to the input sets.

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9 Appendix

9.1 Mathematics

9.1.1 Equations & Derivations

9.2 Ordinary Least Squares (OLS)

The OLS regression is the most prominent statistical model in asset pricing theory. Rosenfeld (2021) summarises OLS. The composition of the true OLS (27) model includes four components. Firstly, \mathbf{X} , an n x k matrix of k independent variables for n observations. Secondly, \mathbf{y} , an n x 1 vector of observation on the dependent variable. Thirdly, ϵ , an n x 1 vector of unexplained error. Lastly, θ , a k x 1 vector of parameters to be estimated.

$$y = X\theta + \epsilon \tag{27}$$

9.2.1 Estimation Criteria

The criteria to obtain the parameter estimate $(\hat{\theta})$ relies on the minimisation of the sum of squared residuals (28). We highlight the observed residuals (e) are distinct from unexplained disturbances (ϵ) . Equation 29 derives residuals by taking the difference between observations based on parameter estimates.

$$\sum e_i^2 \tag{28}$$

$$e = y - X\hat{\theta} \tag{29}$$

Expanding the quadratic $e^T e$ after substituting in equation 29 leads to the alternative expression of the sum of squared residuals in equation 30. Minimizing the sum of square residuals requires taking the partial derivative of equation 30 with respect to the estimated parameters (equation) using matrix differentiation (31). It is imperative X has full rank where all vectors in the matrix are linearly independent, validating both the presence of

a positive definite matrix and minimum.

$$e^T e = y^T y - 2\hat{\theta}^T X^T y + \hat{\theta}^T X^T \hat{\theta} X \tag{30}$$

$$\frac{\partial e^T e}{\partial \hat{\theta}} = -2X^T y + 2X^T X \hat{\theta} = 0 \tag{31}$$

We find the expression for the Ordinary Least Squares (OLS) estimator (32) after rearranging equation 31 to normal form, utilizing inverse matrices to form identity matrices, and simplifying.

$$2X^{T}X\hat{\theta} = 2X^{T}y$$

$$(X^{T}X)^{-1}(X^{T}X)\hat{\theta} = (X^{T}X)^{-1}X^{T}y$$

$$I\hat{\theta} = (X^{T}X)^{-1}X^{T}y$$

$$\hat{\theta} = (X^{T}X)^{-1}(X^{T}y)$$
(32)

Therefore, we can use the OLS estimator to make predictions with OLS (33).

$$\hat{y} = X^T \hat{\theta} \tag{33}$$

9.2.2 Properties of OLS Estimators

There are six key properties in addition to the satisfaction in minimizing the summation of squared residuals.

- 1. The residuals are uncorrelated with the observed values of X i.e., $X^T e = 0$.
- 2. The sum of the residuals is zero i.e., $\sum e_i = 0$.
- 3. The sample mean of the residuals is zero i.e., $\bar{e} = \frac{\sum e_i}{n} = 0$.
- 4. The regression hyperplane passes through the means of observed values i.e., $\frac{e}{=} \frac{y X\theta}{n} = 0$. Since $\bar{e} = 0$ assumed, it is implied $\bar{y} = \bar{x}\bar{\theta}$.

- 5. The residuals are uncorrelated with the predicted y i.e., $\hat{y} = X\hat{\theta}$, $\hat{y}^T e = (X\hat{\beta})^T e = b^T X^T e = 0$
- 6. The mean of \hat{y} for the sample will equal the mean of the y.

9.2.3 The Gauss-Markov Theorem

However, OLS makes Gauss-Markov assumptions about the true model to make inferences regarding β from $\hat{\beta}$. The intention of the Gauss-Markov Theorem, conditional on the below assumptions, states the OLS estimator is the best linear, unbiased, and efficient estimator:

$$y = x\beta + \epsilon$$

$$E[\epsilon|X] = 0$$
 (34)
$$E(\epsilon \epsilon^T | X) = \Omega = \sigma^2 I$$
 (35)
$$\epsilon | X N[0, \sigma^T I] \text{ (hypothesis testing)}$$

- X is an n x k matrix of full rank
- X must be generated randomly, or fixed, by a mechanism uncorrelated to disturbances.

Equation 34 implies $E(y) = X\beta$ as no observations of the independent variables convey any information about the expected values of the disturbances. Equation 35 captures homoskedasticity and no autocorrelation assumptions.

9.3 Tensorflow Model Performance

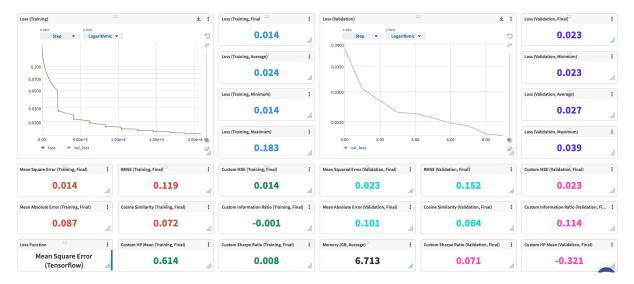


Figure 4: Mean Square Error Model (Tensorflow)

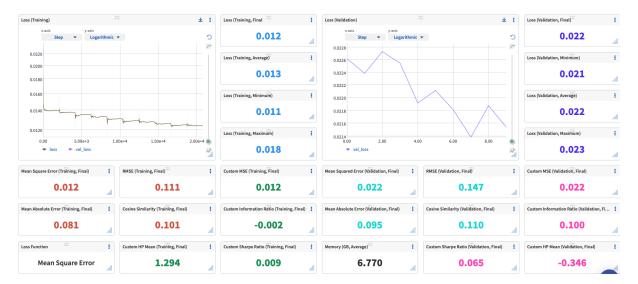


Figure 5: Mean Square Error Loss Function

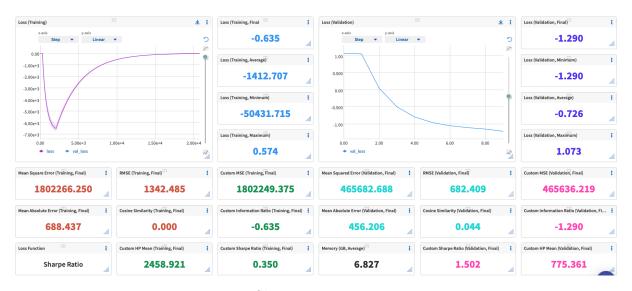


Figure 6: Sharpe Ratio Loss Function

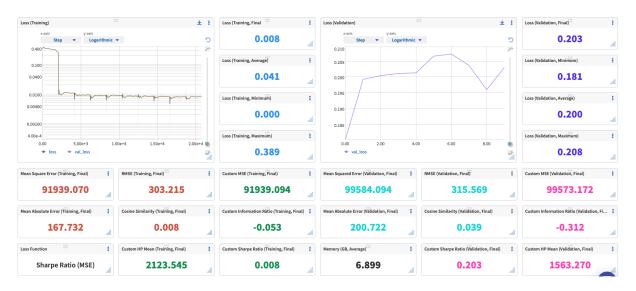


Figure 7: Sharpe Ratio (MSE) Loss Function

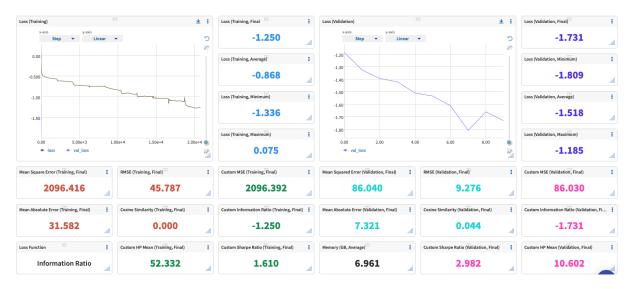


Figure 8: Information Ratio Loss Function

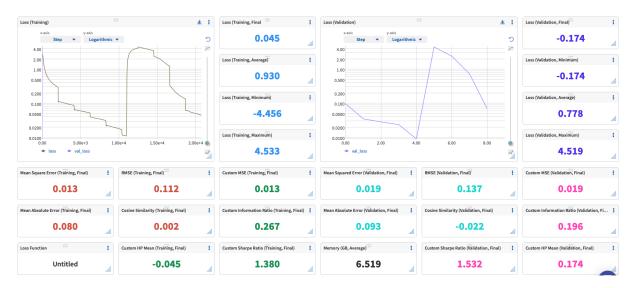


Figure 9: Hedge Portfolio Loss Function

9.4 Pooled Ordinary Least Square Regressions

9.4.1 In-Built Mean Square Error

Dep. Variable:	ret_exc_lead1i	m R	-squared:		0.0019
Estimator:	PooledOLS	\mathbf{R}	-squared ((Between):	0.0038
No. Observations:	531461	\mathbf{R}	-squared ((Within):	0.0011
Date:	Wed, Jan 12 20	022 R	-squared ((Overall):	0.0019
Time:	05:51:31	Lo	og-likeliho	od	3.012e+05
Cov. Estimator:	Clustered				
		F-	statistic:		1001.4
Entities:	7282	P	-value		0.0000
Avg Obs:	72.983	D	istribution	ı:	F(1,531460)
Min Obs:	1.0000				
Max Obs:	252.00	F-	statistic (robust):	13.063
		P	-value		0.0003
Time periods:	252	D	istribution	ı:	F(1,531460)
Avg Obs:	2109.0				
Min Obs:	1869.0				
Max Obs:	3400.0				
Paramete	er Std. Err.	T-stat	P-value	Lower CI	Upper CI
predict 0.0332	0.0092	3.6143	0.0003	0.0152	0.0513

9.4.2 Custom Mean Square Error

Dep. Variable:	ret_exc_lead1m	R-squared:	0.0029
Estimator:	PooledOLS	R-squared (Between):	0.0035
No. Observations:	531461	R-squared (Within):	0.0030
Date:	Wed, Jan 12 2022	R-squared (Overall):	0.0029
Time:	05:52:03	Log-likelihood	3.015e + 05
Cov. Estimator:	Clustered		
		F-statistic:	1564.3
Entities:	7282	P-value	0.0000
Avg Obs:	72.983	Distribution:	F(1,531460)
Min Obs:	1.0000		
Max Obs:	252.00	F-statistic (robust):	6.7467
		P-value	0.0094
Time periods:	252	Distribution:	F(1,531460)
Avg Obs:	2109.0		
Min Obs:	1869.0		
Max Obs:	3400.0		
Paramet	er Std. Err. T-	stat P-value Lower CI	Upper CI
predict 0.0449	0.0173 2.5	5974 0.0094 0.0110	0.0788

9.4.3 Custom Hedge Portfolio

Dep. Variable:	ret_exc_lead1ı	n R-squared:	0.0007
Estimator:	PooledOLS	R-squared (Between):	-0.0026
No. Observations:	531461	R-squared (Within):	0.0010
Date:	Wed, Jan 12 20	022 R-squared (Overall):	0.0007
Time:	05:54:08	Log-likelihood	3.009e + 05
Cov. Estimator:	Clustered		
		F-statistic:	358.78
Entities:	7282	P-value	0.0000
Avg Obs:	72.983	Distribution:	F(1,531460)
Min Obs:	1.0000		
Max Obs:	252.00	F-statistic (robust):	25.217
		P-value	0.0000
Time periods:	252	Distribution:	F(1,531460)
Avg Obs:	2109.0		
Min Obs:	1869.0		
Max Obs:	3400.0		
Paramete	er Std. Err.	T-stat P-value Lower CI	Upper CI
predict 0.0689	0.0137	5.0216 0.0000 0.0420	0.0959

9.5 Technical Details

9.5.1 Organisation

This research essay uses data science best practise (Wilson et al., 2016). Data and results saved regularly and reproducible. Data retention in all forms receives high levels of attention. Project files synchonises continuously to Google Drive (Google LLC, 2020). Git (Linus Torvalds, 2020) manages version control protocols for source code, data, documents, and results. Git stores a complete history of versions using Git hashes. These hashes are strings unique to each state of the publicly available finance-honours repos-

itory¹. Git hashes enable discretisation of finance-honours development, enabling the accessibility and recollection of all previous states given a unique git hash. This functionality enables reproducibility, error correction, and the ability to revert to previous models.

9.5.2 Version Control

Git, hosted by GitHub, provides a comprehensive set of version control technologies and range of benefits. These technologies manage version control for the the programming of approximately 40 methods, classes, and functions. Firstly, Git enables collaborative functionalities. The master version of a project is accessible for all who have access to the repository. Each contributor can create custom copies of branches through pull requests on the master branch. Contributors can commit changes to custom branches and push these changes to the master branch through push requests. Product managers can review push requests, approving valid requests for integrating changes to the master branch. Collaborative efforts are possible with commit messages describing contributions from each contributor. This research essay has only one contributor, rendering collaborative functionalities redundant in this instance. Git ensures the storage of code, work, and author histories. The descriptive nature of commit logs ensures journal accuracy.

9.5.3 Directories

This research essay follows directory structure recommendations from Wilson et al (2016). Organisation is crucial as the modelling of artificial neural networks involves integrating a range of optimisation models, data files and documents. Directory management is most efficient and comprehensive. **finance-honours** is the root directory containing the following sub directories: bin, data, doc, src, and results. The **bin** sub directory contains external scripts and compiled programmes. The **data** sub directory contains all raw data associated with the project. The **doc** sub directory stores user guides, academic resources, research reports and project deliverables. The **results** sub directory contains

¹https://github.com/CMCD1996/finance-honours

the outputs from project analysis. The **src** sub directory stores the source code for preparing datasets, partitioning sets of geographies with varying granularities. All files were continuously backed up using Google Drive and Git.

9.5.4 Python

Python 3.9.7 is the primary programming language for this research essay. The language is omnipresent, widespread in software development. Python's language design makes the language highly productive and simple to use. Python can hand off computationally straining tasks to C/C++ using supporting first-class integration capabilities. The language also has a very active and supportive community. Python is the most popular coding language on the planet defined by the PYPL PopularitY of Programming Language Index. As at December 2021, Python has 30.21% of all language tutorial search instances on Google (PYPL, 2021). Python's dynamic, low cost, and open source nature makes programming quick.

9.5.5 Package Management

The Anaconda package management platform for Python (Anaconda, Inc., 2020) is the chosen coding environment. Anaconda is a well defined, free platform, with known versions of python packages such as matplotlib, numpy, and pip. The use of this environment ensures reproducibility and consistency across infrastructure. Pip is the default package manager for Python, included in the Anaconda package. Pip manages package installation and updates.

9.5.6 Code Style

The PEP8 style for Python Code is formatting style for development code Guido Van Rossum and Coghlan, 2001. Yapf, a formatter maintained by Google, manages formatting. Standardised formatting is important as makes supports readability, optimisation, and consistency. Docstrings and rigourous commenting are important in documentation. A docstring is a Python inline comment describing function use, inputs, and outputs.

An unique docstring belongs to each Python class and function. The Google style docstring is most appropriate because of it's readability, writing ease, and consistency with Google's Style Guide. The parsing of yapf docstrings enables automated documentation generators to create docstring documents describing functions and classes.

9.5.7 Infrastructure

This research essay deploys variations in artificial neural networks of changing size and complexity. Analysis either took place locally, or remotely, depending on the computational requirements for the particular analysis. An Apple MacBook Pro 13 Inch 2019 with 8 GB 2133 MHz LPDDR3 memory and 1.4 GHz Quad-Core Intel Core i5 processor handles simple tasks locally. A Virtual Machine Instance on the Google Cloud Platform Insert specification before submission handles complex tasks remotely.

9.5.8 Documentation

The research essay documentation keeps an accurate record of key design decisions. Commit histories (9.5.2) is the most important form of documentation. Application of auxiliary documentation methods are supplementary.

9.6 Code

All files, resources, and code is available for download from Github. The document listing function and class docstring is available for download here. Furthermore, the coding listings for this research essay follow. Try update.

9.7 Data

9.7.1 Composition & Sources

The authors provide documentation and web-based resources on GitHub to reconstruct an updated dataset from Wharton Research Data Services (WRDS) using SAS Studio. Identifier variables (e.g., size group), Accounting variables (e.g., COGS), accounting characteristics (e.g., change in net working capital, solvency ratios etc.), market variables (e.g., share price, excess return), market characteristics (e.g., market equity, 60 month CAPM β), and detailed characteristics (e.g., equity duration, Altman Z-Score) and Foreign Exchange Conversion Rates feature in the dataset's composition.

9.7.2 Data Processing

However, the computational complexity exceeds resources available at the time of analysis. The replacement of NaN values in a feature columns with the median value of the respective column to retain observations. Subsequently, training sets require further preprocessing in addition to reconfiguring infrastructure. Furthermore, the reduction in numerical feature precision from float64 to float32 effectively halves memory usage. Memory monitoring methods accompany the aforementioned preprocessing adjustments, monitoring CPU and GPU utilisation, reconfiguring GPU's, and configuring application programming interfaces for monitoring modelling performance.

9.7.3 Summary Statistics

The dataset is exhaustive as illustrated by the both summary statistics and Global Factor Data Documentation in the author's GitHub repository. Table 4 describes summary statistics for the entire global factor dataset. Tables 5, 6, and 7 list summary statistics for revised training, validation, and testing sets, respectively.

	count	mean	std	min	25%	20%	%92	max
permno	2739928.0		5.405281e + 04 2.782267e + 04 10000.0000	10000.0000	2.651800e+04	5.715400e+04	5.715400e + 04 8.018600e + 04 9.343600e + 04	9.343600e+04
permco	2739928.0	1.843974e+04 1	1.402881e+04	3.0000	7.702000e+03	1.640850e+04	1.640850e + 04 2.321000e + 04 5.766700e + 04	5.766700e+04
crsp_shrcd	2739928.0	1.089520e+01	4.571000e-01	10.0000	1.1000000e+01	1.1000000e+01	1.100000e+01 1.100000e+01	1.200000e+01
crsp_exchcd	2739928.0	2.127400e+00	9.343000e-01	1.0000	1.000000e+00	3.0000000e+00	3.000000e+00 3.000000e+00 3.000000e+00	3.0000000e+00
sic	2692217.0	4.605936e+03	1.921398e+03	100.0000	3.271000e+03	4.011000e+03	4.011000e + 03 6.036000e + 03 9.999000e + 03	9.9990000e+03
ff49	2674304.0	3.037380e+01	1.341740e+01	1.0000	1.800000e+01	3.4000000e+01	3.400000e+01 4.300000e+01	4.900000e+01
adjfct	2739928.0	2.838700e+00	1.267170e+01	0.0000	1.000000e+00	1.0000000e+00	1.0000000e+00 2.0000000e+00	1.215000e+03
shares	2739928.0	6.078630e+01	2.852566e + 02	0.0830	4.399000e+00	1.251900e+01	.251900e+01 3.808200e+01	2.920640e+04
me	2739928.0	2.241254e+03	1.473073e+04	1.1708	4.367020e+01	1.565628e+02	1.565628e + 02 7.167608e + 02	2.255969e+06
me_company	2739928.0	2.283180e+03	1.527340e+04	1.1708	4.387450e+01	1.574086e+02	1.574086e + 02 7.211363e + 02	2.255969e+06
prc	2739928.0	2.876220e+01	6.488772e+02	0.0078	7.875000e+00	1.612500e+01	1.612500e + 01 2.912500e + 01	1.416000e+05
prc_local	2739928.0	2.876220e+01	6.488772e+02	0.0078	7.875000e+00	1.612500e+01	1.612500e + 01 2.912500e + 01	1.416000e+05
dolvol	2580622.0	3.282292e+08	2.520900e+09	0.0000	1.070786e + 06	7.165154e+06	7.076108e+07	8.441730e+11
ret	2719460.0	1.640000e-02	1.672000e-01	-1.0000	-5.880000e-02	4.100000e-03	7.410000e-02	2.4000000e+01
ret_local	2719460.0	1.640000e-02	1.672000e-01	-1.0000	-5.880000e-02	4.100000e-03	7.410000e-02	2.4000000e+01
ret_exc	2719460.0	1.270000e-02	1.673000e-01	-1.0068	-6.250000e-02	7.000000e-04	7.060000e-02	2.399690e+01
ret_lag_dif	2739928.0	1.0000000e+00	0.0000000e+00	1.0000	1.000000e+00	1.0000000e+00	1.0000000e+00	1.0000000e+00
ret_exc_lead1m	2732542.0	6.400000e-03	1.559000e-01	-1.0113	-6.560000e-02	-1.800000e-03	6.710000e-02	1.988170e+01
market_equity_rank_x	2739928.0	5.982920e+01	2.380660e+01	1.0000	4.000000e+01	6.0000000e+01	8.000000e+01	9.9500000e+01
enterprise_value_rank_x 2480615.0	2480615.0	5.845440e+01	2.501660e+01	1.0000	3.800000e+01	5.900000e+01 8.000000e+01		9.950000e+01
book_equity_rank_x	2452453.0		2.593820e+01	1.0000	3.800000e+01	5.9000000e+01	8.0000000e+01	9.950000e+01
assets_rank_x	2522907.0	5.751850e+01	2.635510e+01	1.0000	3.700000e+01	5.900000e+01	8.0000000e+01	9.950000e+01
sales_rank_x	2509790.0	5.691950e+01	2.717080e+01	1.0000	3.600000e+01	5.900000e+01	5.900000e+01 8.000000e+01	9.950000e+01
net_income_rank_x	2517298.0	5.581200e+01	2.878360e+01	1.0000	3.300000e+01	6.000000e+01 8.000000e+01		9.9500000e+01
bidask_x	2739928.0	1.289000e-01	3.351000e-01	0.0000	0.000000e+00	0.0000000e+00	0.0000000e+00 0.0000000e+00	1.0000000e+00
prc_high_x	2355383.0		2.608370e+01	0.1790	9.250000e+00	1.850000e+01	1.850000e+01 3.300000e+01	4.617600e + 02
prc_low_x	2365005.0		2.325750e+01	0.0818	7.640000e+00	1.6000000e+01		4.175300e+02
tvol_x	2580622.0	∞	2.941295e+07	0.0000	9.875000e+04	5.5100000e+05		6.485186e + 08
$\operatorname{div1m.me.x}$	2718102.0	1.300000e-03	3.700000e-03	0.0000	0.0000000e+00	0.0000000+000	0.0000000e+00 0.0000000e+00	9.0100000e-02

	count	mean	std	min	25%	%02	\%22	max
div3m_me_x	2718121.0	4.000000e-03	6.0000000e-03	0.0000	0.0000000+00	0.00000000+000	6.700000e-03	1.164000e-01
div6m_me_x	2660395.0	8.100000e-03	1.170000e-02	0.0000	0.000000e+00	0.0000000e+00	1.360000e-02	1.472000e-01
div12m_me_x	2548844.0	1.670000e-02	2.350000e-02	0.0000	0.0000000+00	3.800000e-03	2.780000e-02	4.015000e-01
$chcsho_1m_x$	2720001.0	3.200000e-03	2.550000e-02	-0.1168	0.000000e+00	0.0000000e+00	0.0000000e+00	1.096800e+00
chcsho_3m_x	2681179.0	1.240000e-02	6.180000e-02	-0.1424	0.000000e+00	0.0000000e+00	3.300000e-03	1.686700e+00
chcsho_6m_x	2624125.0	2.810000e-02	1.189000e-01	-0.1880	0.00000000+00	9.000000e-04	1.070000e-02	3.832600e+00
$chcsho_12m_x$	2514147.0	6.190000e-02	2.297000e-01	-0.2696	0.000000e+00	4.700000e-03	3.390000e-02	8.477000e+00
eqnpo_1m_x	2718435.0	-1.500000e-03	2.310000e-02	-0.6801	-0.0000000e+00	0.0000000e+00	0.0000000e+00	1.263000e-01
eqnpo_3m_x	2677912.0	-6.200000e-03	5.200000e-02	-0.9973	-1.800000e-03	0.0000000e+00	8.000000e-03	1.696000e-01
eqnpo_6m_x	2618619.0	-1.350000e-02	8.900000e-02	-1.5754	-7.400000e-03	0.0000000e+00	1.640000e-02	2.788000e-01
eqnpo_12m_x	2504936.0	-2.670000e-02	1.474000e-01	-2.2489	-2.450000e-02	0.0000000e+00	3.340000e-02	4.743000e-01
$ret_1_0_x$	2541516.0	1.490000e-02	1.481000e-01	-0.7242	-6.120000e-02	7.900000e-03	7.690000e-02	2.176500e+00
ret_2_0x	2521767.0	2.960000e-02	2.125000e-01	-0.8327	-8.110000e-02	1.480000e-02	1.176000e-01	3.342500e+00
ret_3_0x	2503682.0	4.400000e-02	2.649000e-01	-0.8864	-9.610000e-02	2.270000e-02	1.506000e-01	5.0000000e+00
ret_3_1_x	2502019.0	2.870000e-02	2.108000e-01	-0.8310	-8.140000e-02	1.440000e-02	1.167000e-01	3.342500e+00
ret_6_0_x	2447794.0	8.830000e-02	3.970000e-01	-0.9396	-1.267000e-01	4.500000e-02	2.336000e-01	8.555600e+00
ret_6_1x	2446030.0	7.230000e-02	3.553000e-01	-0.9171	-1.184000e-01	3.700000e-02	2.059000e-01	8.411800e+00
ret_9_0_x	2393988.0	1.336000e-01	5.093000e-01	-0.9721	-1.466000e-01	6.750000e-02	3.069000e-01	9.857100e+00
$ret_{-9-1.x}$	2392087.0	1.168000e-01	4.700000e-01	-0.9555	-1.414000e-01	5.930000e-02	2.812000e-01	9.273700e+00
ret_12_0_x	2341375.0	1.813000e-01	6.179000e-01	-0.9783	-1.593000e-01	9.080000e-02	3.773000e-01	1.301590e+01
ret_12_1x	2339380.0	1.635000e-01	5.789000e-01	-0.9728	-1.558000e-01	8.200000e-02	3.514000e-01	1.223080e+01
ret_12_7x	2337747.0	7.050000e-02	3.478000e-01	-0.9055	-1.163000e-01	3.610000e-02	2.015000e-01	8.509400e+00
ret_18_1x	2239551.0	2.625000e-01	7.812000e-01	-0.9850	-1.710000e-01	1.321000e-01	4.926000e-01	2.048480e+01
ret_24_1x	2145964.0	3.596000e-01	9.260000e-01	-0.9890	-1.717000e-01	1.837000e-01	6.267000e-01	1.484620e+01
$ret_24_12_x$	2142652.0	1.821000e-01	6.037000e-01	-0.9678	-1.493000e-01	9.260000e-02	3.714000e-01	1.345160e+01
ret_36_1x	1976435.0	5.673000e-01	1.234400e+00	-0.9935	-1.548000e-01	2.964000e-01	8.916000e-01	1.914000e+01
$ret_36_12_x$	1972590.0	3.838000e-01	9.482000e-01	-0.9864	-1.546000e-01	2.006000e-01		1.702520e+01
ret_48_12_x	1821582.0		1.256400e+00	-0.9918	-1.358000e-01	3.161000e-01	9.172000e-01	1.811810e+01
ret_48_1_x	1826053.0	7.976000e-01	1.577300e+00	-0.9965	-1.285000e-01	4.175000e-01	4.175000e - 01 1.176300e + 00 1.772000e + 01	1.772000e+01

	count	mean	std	min	25%	20%	12%	max
ret_60_1x	1691563.0	1.064400e+00	2.014800e+00	-0.9985	-9.170000e-02	5.486000e-01	5.486000e-01 1.492300e+00 2.754720e+01	2.754720e+01
$ret_60_12_x$	1686573.0	8.258000e-01	1.611700e+00	-0.9960	-1.096000e-01	4.364000e-01	1.200000e + 00 2.063640e + 01	2.063640e+01
$ret_60_36_x$	1680619.0	3.857000e-01	9.340000e-01	-0.9860	-1.429000e-01	2.072000e-01	6.479000e-01	1.808570e+01
seas_1_1an_x	2426517.0	1.420000e-02	1.421000e-01	-0.6705	-6.040000e-02	7.600000e-03	7.560000e-02	1.823500e+00
seas_1_1na_x	1870192.0	1.490000e-02	4.360000e-02	-0.2355	-7.800000e-03	1.280000e-02	3.460000e-02	3.871000e-01
seas_2_5an_x	1599992.0	1.520000e-02	6.790000e-02	-0.2970	-2.260000e-02	1.180000e-02	4.810000e-02	6.337000e-01
at_gr1_x	2426455.0	2.641000e-01	9.239000e-01	-0.7398	4.800000e-03	9.050000e-02	2.391000e-01	3.163840e+01
ca_gr1_x	2184566.0	3.206000e-01	1.336600e+00	-0.8313	-3.830000e-02	9.400000e-02	2.815000e-01	4.636900e+01
nca_gr1_x	2183067.0	3.950000e-01	1.682300e+00	-0.8737	-1.530000e-02	8.250000e-02	2.844000e-01	5.781320e+01
$ltgr1_x$	2408077.0	3.042000e-01	9.791000e-01	-0.8021	-2.990000e-02	8.560000e-02	2.894000e-01	1.783760e+01
cl_gr1_x	2190296.0	2.996000e-01	8.898000e-01	-0.8494	-6.490000e-02	1.114000e-01	3.701000e-01	1.634630e+01
ncl_gr1_x	2075342.0	9.926000e-01	5.509500e+00	-1.0000	-1.023000e-01	3.970000e-02	3.376000e-01	1.990000e+02
be_gr1_x	2311345.0	3.178000e-01	1.301000e+00	-0.9166	5.900000e-03	9.660000e-02	2.271000e-01	3.373330e+01
$debt_gr1_x$	2158693.0	7.838000e-01	4.707200e+00	-1.0000	-1.456000e-01	1.900000e-02	3.292000e-01	1.090000e+02
sale_gr1_x	2362404.0	2.228000e-01	6.711000e-01	-0.9960	5.000000e-03	1.032000e-01	2.478000e-01	1.370570e+01
$\cos - \operatorname{gr}_{-x}$	2358805.0	2.142000e-01	6.122000e-01	-0.9619	-4.700000e-03	1.032000e-01	2.613000e-01	1.190030e+01
sga_gr1_x	1997437.0	1.844000e-01	3.963000e-01	-1.0000	1.340000e-02	1.044000e-01	2.389000e-01	6.765800e+00
opex_gr1_x	2387208.0	1.949000e-01	4.470000e-01	-0.7668	7.900000e-03	1.058000e-01	2.505000e-01	7.187400e+00
capx_gr1_x	2147147.0	6.016000e-01	2.183000e+00	-1.3370	-2.236000e-01	1.144000e-01	6.251000e-01	3.425000e+01
inv-gr1_x	1910333.0	2.595000e-01	9.931000e-01	-1.0000	-6.850000e-02	8.260000e-02	2.909000e-01	1.698080e+01
at_gr3_x	2114339.0	9.104000e-01	2.670800e+00	-0.8797	8.870000e-02	3.426000e-01	8.167000e-01	6.899070e+01
ca_gr3_x	1898998.0	9.832000e-01	3.187300e+00	-0.9099	2.890000e-02	3.230000e-01	8.289000e-01	7.748590e+01
nca_gr3_x	1897746.0	1.592100e+00	6.786800e+00	-0.9628	4.280000e-02	3.455000e-01	1.005000e+00	1.792615e+02
lt_gr3_x	2091277.0	1.135900e+00	3.376000e+00	-0.8936	3.580000e-02	3.474000e-01	9.457000e-01	5.633890e+01
cl_gr3_x	1906078.0	9.845000e-01	2.656400e+00	-0.9194	9.000000e-03	3.652000e-01	9.754000e-01	4.535460e+01
ncl_gr3_x	1803330.0	4.168200e+00	2.242620e+01	-1.0000	-1.231000e-01	2.914000e-01	1.285200e+00	8.323333e+02
be_gr3_x	1998122.0	1.009400e+00	3.275200e+00	-0.9384	7.210000e-02	3.326000e-01	7.902000e-01	6.699660e+01
$debt_gr3_x$	1882647.0	3.622500e+00	2.086590e+01	-1.0000	-2.165000e-01	2.251000e-01	1.145100e+00 4.310000e+02	4.3100000e+02
sale_gr3_x	2063618.0	8.605000e-01	2.814400e+00	-1.0000	7.210000e-02	3.286000e-01	7.527000e-01	7.527000e-01 8.620390e+01

$\frac{\text{count}}{2052669.0 }$
1.692700e + 00 5.902400e + 00 1.480000e - 02 1.380000e - 01
1.250000e-02 5.090000e-02
2.190000e-02 6.430000e-02
5.400000e-03 $4.060000e-02$
$3.900000e-03 \mid 6.220000e-02$
$1.460000e$ -02 $\mid 4.890000e$ -02
$9.000000e$ -04 $\mid 1.130000e$ -02
$1.770000 \mathrm{e} \text{-} 02 \ 9.970000 \mathrm{e} \text{-} 02$
2.300000e-03 1.280000e-02
$3.450000 \mathrm{e} ext{-}02 \Big \ 1.005000 \mathrm{e} ext{-}01$
1.980000e-02 6.480000e-02
$1.440000e$ -02 $\mid 8.680000e$ -02
$4.890000 e - 02 \Big 1.438000 e - 01$
$6.300000e$ -03 $\mid 3.310000e$ -02
4.270000e-02 1.424000e-01
8.310000e-02 2.025000e-01
$2.620000 \mathrm{e} ext{-}02 8.090000 \mathrm{e} ext{-}02$
5.700000e-03 6.030000e-02
2.150000e- $02 1.353000$ e- 01
-1.580000e- 02 1.552000 e- 01
$3.580000e-02 \Big 1.161000e-01$
$9.700000e$ -03 $\mid 9.740000e$ -02
$5.200000e-03 \ 9.760000e-02$

ope-gr1a.x 2056758.0 9.40 ni-gr1a.x 2402691.0 8.00 nix-gr1a.x 2309627.0 3.90 dp-gr1a.x 2334713.0 1.00 fcf_gr1a.x 2181931.0 -7.30 nwc_gr1a.x 2181931.0 -7.30 nwc_gr1a.x 2164316.0 2.64 eqnetis.gr1a.x 2052797.0 1.17 dltnetis.gr1a.x 2290818.0 7.00 eqnpo-gr1a.x 2047069.0 -1.04 tax_gr1a.x 2047069.0 1.70 eqbb-gr1a.x 2398103.0 3.10 eqsb-gr1a.x 2000469.0 1.36 eqs.gr1a.x 2000469.0 1.30 eqs.gr1a.x 2000469.0 1.30 eqs.gr1a.x 2000469.0 2.90 capx_gr1a.x 2382722.0 1.10 be-gr1a.x 2311289.0 2.90 capx_gr1a.x 2033267.0 2.90 inv-gr3a.x 2047864.0 2.90 ppee_gr3a.x 2047864.0 2.29		-0.9869 -1.6889 -1.8549 -0.3935 -2.0255 -0.9941 -1.1368 -1.4272 -1.9975 -0.7874 -0.7874 -0.7874 -0.7874	-0.9869 -1.390000e-02 -1.6889 -1.340000e-02 -1.8549 -1.540000e-02 -2.0255 -5.480000e-02 -0.9941 -4.190000e-02 -1.1368 -6.050000e-02 -1.4272 -2.650000e-02 -1.9975 -1.00000e-02 -0.7874 -2.580000e-02 -0.7874 -2.580000e-02 -1.0269 -4.130000e-02 -1.0269 -4.130000e-02 -1.0269 -4.130000e-02 -1.0269 -4.130000e-02 -1.0269 -4.130000e-02	1.090000e-02 3.900000e-03 3.800000e-03 2.500000e-03 2.900000e-04 1.650000e-02 0.000000e+00 0.000000e+00 0.000000e+00 1.700000e-03	3.950000e-02 2.570000e-02 7.500000e-03 7.330000e-02 4.640000e-02 5.020000e-02 7.240000e-02 1.380000e-02 1.380000e-02 4.330000e-02 7.550000e-02	3.950000e-02 1.233300e+00 2.430000e-02 2.739400e+00 2.570000e-02 2.791300e+00 7.500000e-03 1.932000e-01 7.330000e-02 1.485100e+00 4.640000e-02 1.151200e+00 7.240000e-02 1.202900e+00 7.240000e-02 1.207600e+00 1.380000e-02 7.003000e-01 1.870000e-02 7.197000e-01 1.870000e-02 7.197000e-01 1.870000e-02 7.197000e-01 1.870000e-02 7.197000e-01 1.870000e-01 7.197000e-01 7.197000e-01 1.870000e-01 7.197000e-01 7.1970
2402691.0 2402691.0 2309627.0 2309627.0 2053075.0 2334713.0 2181931.0 -2164316.0 2164316.0 2373431.0 -2290818.0 -2374474.0 -2374474.0 -2373431.0 -2373431.0 -2398103.0 2382722.0 2382722.0 2382722.0 2382722.0 2382722.0 2382722.0 2382722.0 2382722.0 2382722.0 2382722.0 2382722.0 2382722.0 2382722.0 2382722.0 2382722.0 2382722.0 2382722.0 2382722.0 2484434.0 2582722.0 2681646.0 26847864.0 2864897.0		-1.6889 -1.8549 -0.3935 -2.0255 -0.9941 -1.1368 -1.4272 -1.9975 -0.7874 -0.8063 -1.0269 -2.0764 -1.1821	-1.340000e-02 -1.540000e-02 0.000000e+00 -5.480000e-02 -4.190000e-02 -2.650000e-02 -1.000000e-02 -2.580000e-02 -1.090000e-02 -4.130000e-02 -4.130000e-02 -4.130000e-02	3.900000e-03 3.800000e-03 2.500000e-03 2.900000e-03 -4.000000e-04 1.650000e-02 0.000000e+00 0.000000e+00 0.000000e+00 1.700000e-03	2.430000e-02 2.570000e-02 7.500000e-03 7.330000e-02 4.640000e-02 5.020000e-02 7.240000e-02 1.380000e-02 2.250000e-02 4.330000e-02 7.550000e-02	2.739400e+00 2.791300e+00 1.932000e-01 1.485100e+00 1.151200e+00 9.090000e-01 1.207600e+00 7.003000e-01 7.197000e-01
2402691.0 2309627.0 2309627.0 2053075.0 2334713.0 2181931.0 2164316.0 2164316.0 2164316.0 2164316.0 2374474.0 2374474.0 2374474.0 2398103.0 2398103.0 2382722.0 2382722.0 2382722.0 2382722.0 2382722.0 2382722.0 2382722.0 2382722.0 2382722.0 2382722.0 2382722.0 2382722.0 2382722.0 2184434.0 2033267.0 2047864.0 2864897.0		-1.8549 -0.3935 -2.0255 -0.9941 -1.1368 -1.4272 -1.9975 -0.7874 -0.8063 -1.0269 -2.0764	-1.540000e-02 0.0000000e+00 -5.480000e-02 -4.190000e-02 -2.650000e-02 -1.000000e-02 -1.090000e-02 -1.130000e-02 -4.130000e-02 -4.130000e-02 -4.130000e-02	3.800000e-03 2.500000e-03 2.700000e-03 -4.000000e-04 1.650000e-02 0.000000e+00 0.000000e+00 0.000000e+00 1.700000e-03		
2309627.0 2053075.0 2053075.0 2181931.0 218431.0 218431.0 2290818.0 2374474.0 2052412.0 2052412.0 2398103.0 2398103.0 2382722.0 2884434.0 2184434.0 2184434.0 2184434.0 2184434.0 2184434.0 2184434.0 2184434.0 2184434.0 2184434.0 2184434.0 2184434.0 2184434.0 2184434.0 2184434.0 2184434.0 2184434.0 2184434.0		-0.3935 -2.0255 -0.9941 -1.1368 -1.4272 -1.9975 -0.7874 -0.8063 -1.0269 -2.0764 -1.1821	0.0000000e+00 -5.480000e-02 -4.190000e-02 -6.050000e-02 -1.000000e-02 -2.580000e-02 -2.580000e-02 -1.090000e-02 -1.480000e-02 -4.130000e-02 -4.130000e-02	2.500000e-03 2.700000e-03 2.900000e-04 1.650000e-02 0.000000e+00 0.000000e+00 0.000000e+00 1.700000e-03		1.932000e-01 1.485100e+00 1.151200e+00 9.090000e-01 1.207600e+00 7.003000e-01 7.197000e-01
2053075.0 2334713.0 2181931.0 2164316.0 2164316.0 2164316.0 2373431.0 2374474.0 2052412.0 x 2047069.0 2382722.0 1891334.0 2382722.0	7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7 7	-2.0255 -0.9941 -1.1368 -1.4272 -1.9975 -0.7874 -0.8063 -1.0269 -2.0764 -1.1821	-5.480000e-02 -4.190000e-02 -6.050000e-02 -1.000000e-02 -2.580000e-02 -1.090000e-02 -4.130000e-02 -6.040000e-02	2.700000e-03 2.900000e-03 -4.000000e-04 1.650000e-02 0.000000e+00 0.000000e+00 0.000000e+00 1.700000e-03		1.485100e+00 1.151200e+00 1.202900e+00 9.090000e-01 1.207600e+00 7.003000e-01 7.197000e-01
2334713.0 2181931.0 2181931.0 1a_x 2052797.0 1a_x 2290818.0 2374474.0 2052412.0 3_x 2047069.0 2398103.0 x 2398103.0 x 2184434.0 x 2033267.0 2033267.0 x 2033267.0 x 2033267.0 x 2033267.0 x 2033267.0 x 21890568.0		-0.9941 -1.1368 -1.4272 -1.9975 -0.7874 -0.8063 -1.0269 -2.0764 -1.1821	-4.190000e-02 -6.050000e-02 -2.650000e-02 -1.000000e-02 -2.580000e-02 -1.090000e-02 -4.130000e-02 -6.040000e-02	2.900000e-03 -4.000000e-04 1.650000e-02 0.000000e+00 0.000000e+00 0.000000e+00 1.700000e-03		1.151200e+00 1.202900e+00 9.090000e-01 1.207600e+00 7.003000e-01 7.197000e-01
2181931.0 -2164316.0 1a_X 2052797.0 1a_X 2290818.0 2374474.0 -2374474.0 -2374474.0 -2398103.0 x 2047069.0 x x 2000469.0 x x 2382722.0 x 2081646.0 x 203267.0 203267.0 x 2184434.0		-1.1368 -1.4272 -1.9975 -0.7874 -0.8063 -1.0269 -2.0764 -1.1821	-6.050000e-02 -2.650000e-02 -1.000000e-02 -2.580000e-02 -1.090000e-02 -4.130000e-02 -6.040000e-02	-4.000000e-04 1.650000e-02 0.0000000e+00 0.0000000e+00 0.0000000e+00 1.700000e-03		1.202900e+00 9.090000e-01 1.207600e+00 7.003000e-01 7.197000e-01
ta_x 2052797.0 1a_x 2052797.0 1a_x 2290818.0 1a_x 2290818.0 x 2052412.0 x 2047069.0 x 2047069.0 x 208103.0 x 208103.0 x 2080469.0 x 2081289.0 x 2081646.0 x 2083267.0 2047864.0 x 1890568.0	7 7 7 8 7 7 7	-1.4272 -1.9975 -0.7874 -0.8063 -1.0269 -2.0764 -1.1821	-2.650000e-02 -1.000000e-02 -2.580000e-02 -1.090000e-02 -4.130000e-02 -6.040000e-02	1.650000e-02 0.000000e+00 0.000000e+00 0.000000e+00 1.70000e-03	7.240000e-02 1.380000e-02 2.250000e-02 1.870000e-02 4.330000e-02 7.550000e-02	9.090000e-01 1.207600e+00 7.003000e-01 7.197000e-01
(a_x) 2052797.0 1a_x 2373431.0 1a_x 2290818.0 1a_x 2374474.0 x 2052412.0 x 2047069.0 x 2398103.0 x 1893504.0 x 2382722.0 x 2184434.0 x 2081646.0 203267.0 2047864.0 x 1890568.0 1864897.0 1864897.0	2 1 2 1 2 2	-1.9975 -0.7874 -0.8063 -1.0269 -2.0764 -1.1821	-1.000000e-02 -2.580000e-02 -1.090000e-02 -4.130000e-02 -6.040000e-02 -1.480000e-02	0.0000000e+00 0.0000000e+00 0.0000000e+00 1.700000e-03	1.380000e-02 2.250000e-02 1.870000e-02 4.330000e-02 7.550000e-02	1.207600e+00 7.003000e-01 7.197000e-01
1a_x 2373431.0 1a_x 2290818.0 x 2374474.0 x_x 2052412.0 x 2047069.0 2398103.0 x 2000469.0 x 2082722.0 x 2081434.0 x 2081646.0 x 203267.0 2047864.0 x 1890568.0	- 0 - 0 C	-0.7874 -0.8063 -1.0269 -2.0764 -1.1821	-2.580000e-02 -1.090000e-02 -4.130000e-02 -6.040000e-02 -1.480000e-02	0.0000000e+00 0.0000000e+00 0.000000e+00 1.700000e-03	2.250000e-02 1.870000e-02 4.330000e-02 7.550000e-02	7.003000e-01 7.197000e-01
1a_x 2290818.0 1a_x 2374474.0 x 2052412.0 8_x 2047069.0 x 2398103.0 x 1893504.0 x 2000469.0 x 2081648.0 x 2081646.0 x 203267.0 2047864.0 x 1890568.0	<u> </u>	-0.8063 -1.0269 -2.0764 -1.1821	-1.090000e-02 -4.130000e-02 -6.040000e-02 -1.480000e-02	0.0000000e+00 0.0000000e+00 1.700000e-03	1.870000e-02 4.330000e-02 7.550000e-02	7.197000e-01
la_x 2374474.0 x 2052412.0 8 x 2047069.0 x 1893504.0 2382722.0 x 1891334.0 2 x 2184434.0 2 x 2081646.0 2 203267.0 2 x 2033267.0 2 2047864.0 2 x 1890568.0 1864897.0	7 2 7	-1.0269 -2.0764 -1.1821	-4.130000e-02 -6.040000e-02 -1.480000e-02	0.0000000e+00 1.700000e-03	4.330000e-02 7.550000e-02	_
x 2052412.0 x 2047069.0 x 2398103.0 1893504.0 x 2000469.0 x 2382722.0 x 1891334.0 x 211289.0 x 2033267.0 2047864.0 x 1890568.0 1864897.0	2 2	-2.0764	-6.040000e-02 -1.480000e-02	1.700000e-03	7.550000e-02	1.017900e+00
x 2398103.0	2	-1.1821	-1.480000e-02	00 1 5000000 0		1.539900e+00
x 1893504.0 x 2000469.0 x 2382722.0 x 1891334.0 x 211289.0 x 2031267.0 2047864.0 x 1890568.0		2171	0 0000000	0.0000000e+00	1.310000e-02	1.940900e+00
x 2000469.0 x 2382722.0 x 1891334.0 x 211289.0 x 2031646.0 x 2033267.0 x 1890568.0	3.100000e-03 2.840000e-02	10.717.0-	-3.800000e-03	1.000000e-03	1.140000e-02	2.047000e-01
x 2382722.0 x 1891334.0 x 211289.0 x 2081646.0 x 203267.0 2047864.0 x 1890568.0	1.700000e-03 $3.370000e-02$	-0.3806	0.0000000e+00	0.0000000e+00	3.000000e-04	2.809000e-01
x 1891334.0 x 2184434.0 2311289.0 x 2081646.0 203267.0 2047864.0 x 1890568.0	1.360000e-02 $2.117000e-01$	-2.0255	-2.500000e-03	0.0000000e+00	5.700000e-03	1.226200e+00
x 1891334.0 x 2184434.0 2311289.0 x 2081646.0 203267.0 2047864.0 x 1890568.0 1864897.0	1.100000e-03 $1.270000e-02$	-0.2183	0.0000000e+00	0.0000000e+00	1.200000e-03	2.439000e-01
x 211289.0 x 2081646.0 2033267.0 2047864.0 x 1890568.0 1864897.0	2.900000e-03 $4.380000e-02$	-0.4620	-1.000000e-04	0.00000000+000	4.100000e-03	3.915000e-01
x 2081646.0 2081646.0 203267.0 2047864.0 x 1890568.0 1864897.0	7.400000e-03 $5.440000e-02$	-0.4868	-7.300000e-03	2.300000e-03	1.940000e-02	4.471000e-01
x 2081646.0 2033267.0 2047864.0 x 1890568.0 1864897.0	4.620000e-02 $1.699000e-01$	-2.0718	1.600000e-03	3.510000e-02	8.970000e-02	8.561000e-01
2033267.0 2047864.0 x 1890568.0 1864897.0	2.960000e-02 $1.755000e-01$	-2.5781	-1.260000e-02	9.500000e-03	6.320000e-02	9.052000e-01
x 2047864.0 x 1890568.0 1864897.0	2.900000e-02 8.700000e-02	-0.6971	0.0000000e+00	6.8000000e-03	5.550000e-02	4.115000e-01
x 1890568.0 1864897.0	4.970000e-02 1.082000e-01	-0.7795	1.400000e-03	3.280000e-02	8.960000e-02	4.887000e-01
1864897.0	1.277000e-01 2.118000e-01	-2.1282	3.190000e-02	1.080000e-01	2.163000e-01	9.231000e-01
	1.290000e-02 $ 7.040000e-02 $	-0.6566	0.0000000e+00	0.0000000e+00	8.800000e-03	4.683000e-01
intan_gr3a_x $ 1784074.0 $ 2.52	2.520000e-02 $1.171000e-01$	-1.7938	-0.0000000e+00	0.0000000e+00	2.360000e-02	6.632000e-01
debtst_gr3a_x $ 2078323.0 $ 8.50	8.500000e-03 $7.970000e-02$	-0.8315	-6.500000e-03	3.000000e-04	2.440000e-02	5.514000e-01
1936459.0 3.44	3.440000e-02 $8.510000e-02$	-0.4973	-3.000000e-04	1.600000e-02	4.880000e-02	4.801000e-01

	count	mean	std	min	25%	%02	15%	max
txp_gr3a_x	1751204.0	1.900000e-03	1.400000e-02	9260.0-	-1.200000e-03	0.00000000+000	4.400000e-03	1.079000e-01
$debtlt_gr3a_x$	2098723.0	4.090000e-02	1.579000e-01	-1.1700	-1.120000e-02	1.060000e-02	1.011000e-01	7.496000e-01
txditc_gr3a_x	1843283.0	6.200000e-03	2.480000e-02	-0.2172	0.0000000e+00	0.0000000e+00	1.330000e-02	1.273000e-01
coa_gr3a_x	1880953.0	7.6600000e-02	1.701000e-01	-1.4412	6.100000e-03	6.190000e-02	1.549000e-01	6.791000e-01
col_gr3a_x	1907173.0	4.420000e-02	9.650000e-02	-0.9653	4.300000e-03	3.750000e-02	8.380000e-02	4.559000e-01
cowc_gr3a_x	1861920.0	3.210000e-02	1.338000e-01	-1.0405	-2.130000e-02	2.260000e-02	9.140000e-02	5.604000e-01
ncoa_gr3a_x	1899708.0	1.091000e-01	2.575000e-01	-4.5815	1.230000e-02	1.026000e-01	2.250000e-01	8.112000e-01
ncol_gr3a_x	1887939.0	1.640000e-02	5.970000e-02	-0.5782	-0.0000000e+00	9.0000000e-03	3.080000e-02	4.104000e-01
nncoa_gr3a_x	1861492.0	9.300000e-02	2.474000e-01	-3.9391	1.200000e-03	8.690000e-02	2.030000e-01	8.094000e-01
oa_gr3a_x	1880920.0	1.840000e-01	3.641000e-01	-5.1474	4.560000e-02	2.082000e-01	3.829000e-01	9.247000e-01
ol_gr3a_x	1887939.0	6.020000e-02	1.295000e-01	-1.1795	1.270000e-02	5.900000e-02	1.138000e-01	6.233000e-01
fna_gr3a_x	2302373.0	1.560000e-02	8.920000e-02	-1.1421	0.0000000e+00	0.0000000e+00	0.0000000e+00	7.162000e-01
fnl_gr3a_x	2105333.0	4.560000e-02	2.040000e-01	-1.8999	-1.910000e-02	2.6000000e-02	1.304000e-01	8.753000e-01
nfna_gr3a_x	2105333.0	-3.150000e-02	2.282000e-01	-1.3255	-1.318000e-01	-2.310000e-02	4.440000e-02	2.048000e+00
gp-gr3a_x	2074121.0	7.850000e-02	1.870000e-01	-1.2858	4.200000e-03	5.550000e-02	1.554000e-01	1.274100e+00
ebitda-gr3a_x	2079592.0	2.410000e-02	1.330000e-01	-1.0362	-8.600000e-03	2.410000e-02	7.360000e-02	1.478800e+00
ebit_gr3a_x	2081034.0	1.490000e-02	1.346000e-01	-1.1637	-1.460000e-02	1.620000e-02	6.010000e-02	1.985300e+00
ope-gr3a_x	1772515.0	2.290000e-02	1.350000e-01	-1.1140	-1.410000e-02	2.540000e-02	7.260000e-02	1.382600e+00
ni-gr3a_x	2095331.0	5.500000e-03	1.607000e-01	-2.0040	-1.480000e-02	8.900000e-03	4.110000e-02	3.365400e+00
nix-gr3a_x	2095331.0	5.200000e-03	1.722000e-01	-2.2144	-1.670000e-02	8.800000e-03	4.270000e-02	3.330500e+00
dp -gr3a_x	1998657.0	9.200000e-03	2.780000e-02	-0.6566	5.000000e-04	7.400000e-03	1.760000e-02	3.627000e-01
ocf_gr3a_x	2026157.0	1.030000e-02	1.536000e-01	-0.9623	-3.950000e-02	1.100000e-02	6.680000e-02	1.459300e+00
fcf_gr3a_x	1875380.0	-2.300000e-03	1.806000e-01	-0.9594	-6.520000e-02	3.5000000e-03	6.430000e-02	1.668700e+00
nwc-gr3a_x	1880705.0	5.470000e-02	2.333000e-01	-3.1433	-2.400000e-02	4.470000e-02	1.438000e-01	9.475000e-01
dltnetis_gr3a_x	2057295.0	-7.000000e-03	1.381000e-01	-0.9437	-3.150000e-02	0.0000000e+00	2.360000e-02	8.602000e-01
dstnetis_gr3a_x	1975805.0	-1.000000e-04	7.960000e-02	-0.7776	-1.420000e-02	0.0000000e+00	1.680000e-02	6.541000e-01
dbnetis_gr3a_x	2058325.0	-7.400000e-03	1.681000e-01	-1.2437	-4.610000e-02	0.0000000e+00	4.140000e-02	1.075700e+00
tax_gr3a_x	2090131.0	6.500000e-03	3.6000000e-02	-0.2190	-4.800000e-03	2.700000e-03	1.970000e-02	2.106000e-01
div_gr3a_x	2069485.0	2.200000e-03	1.420000e-02	-0.2110	0.0000000e+00	0.0000000e+00	4.200000e-03	2.609000e-01

	count	mean	std	min	25%	20%	15%	max
capx_gr3a_x	1877910.0	1.340000e-02	6.720000e-02	-0.6838	-6.700000e-03	6.500000e-03	3.240000e-02	3.679000e-01
capx_at_x	2305667.0	6.630000e-02	7.300000e-02	-0.0305	1.920000e-02	4.470000e-02	8.570000e-02	6.092000e-01
spi_at_x	2376699.0	-1.0100000e-02	4.960000e-02	-1.3123	-2.700000e-03	0.0000000e+00	0.0000000e+00	1.961000e-01
xido_at_x	2513016.0	-5.000000e-04	1.800000e-02	-0.4152	0.000000e+00	0.0000000e+00	0.0000000e+00	1.762000e-01
nri_at_x	2375825.0	-1.080000e-02	6.070000e-02	-1.5759	-4.600000e-03	0.0000000e+00	0.0000000e+00	2.675000e-01
gp_sale_x	2468341.0	8.440000e-02	3.062100e+00	-124.7476	2.080000e-01	3.345000e-01	5.045000e-01	9.763000e-01
ebitda_sale_x	2470375.0	-3.073000e-01	4.409900e+00	-171.6176	5.970000e-02	1.272000e-01	2.277000e-01	7.373000e-01
ebit_sale_x	2470818.0	-3.840000e-01	4.578500e+00	-185.0447	3.170000e-02	8.990000e-02	1.721000e-01	6.154000e-01
pi_sale_x	2473639.0	-4.469000e-01	4.876400e+00	-184.2990	1.190000e-02	7.260000e-02	1.445000e-01	7.101000e-01
ni_sale_x	2474362.0	-4.693000e-01	4.796100e+00	-184.2990	7.200000e-03	4.550000e-02	9.440000e-02	5.566000e-01
nix_sale_x	2472905.0	-4.745000e-01	4.848700e+00	-184.2990	6.200000e-03	4.620000e-02	9.640000e-02	6.508000e-01
ocf_sale_x	2414346.0	-3.439000e-01	3.755000e+00	-140.2577	-1.520000e-02	5.800000e-02	1.448000e-01	1.412300e+00
fcf_sale_x	2267091.0	-5.418000e-01	4.134400e+00	-125.9694	-1.053000e-01	-1.100000e-03	6.670000e-02	1.210500e+00
gp_at_x	2503159.0	3.0110000e-01	2.895000e-01	-1.2660	1.023000e-01	2.659000e-01	4.563000e-01	1.412300e+00
ebitda_at_x	2505194.0	7.710000e-02	1.992000e-01	-2.1076	2.950000e-02	1.080000e-01	1.699000e-01	5.122000e-01
ebit_at_x	2506116.0	4.100000e-02	1.986000e-01	-2.1142	1.820000e-02	7.130000e-02	1.269000e-01	4.730000e-01
fi_at_x	2185678.0	1.660000e-02	2.114000e-01	-2.6041	2.010000e-02	6.410000e-02	9.800000e-02	3.716000e-01
cop_at_x	2259456.0	1.333000e-01	1.925000e-01	-1.1882	3.940000e-02	1.365000e-01	2.302000e-01	1.940400e+00
ni_at_x	2514966.0	-5.0000000e-03	2.045000e-01	-2.8828	3.400000e-03	3.510000e-02	7.410000e-02	3.332000e-01
ope-be-x	2108352.0	1.569000e-01	5.427000e-01	-8.8149	9.490000e-02	2.136000e-01	3.261000e-01	3.725100e+00
ni_be_x	2444347.0	-1.990000e-02	5.962000e-01	-10.7541	1.720000e-02	9.500000e-02	1.504000e-01	1.450500e+00
nix_be_x	2444347.0	-2.270000e-02	6.187000e-01	-11.9515	1.490000e-02	9.590000e-02	1.526000e-01	1.558300e+00
ocf_be_x	2375509.0	4.150000e-02	5.350000e-01	-7.2459	-3.990000e-02	1.089000e-01	2.199000e-01	4.068700e+00
fcf_be_x	2219533.0	-1.352000e-01	6.520000e-01	-9.8959	-2.117000e-01	-4.000000e-03	1.206000e-01	2.895100e+00
gp-bev-x	2404319.0	6.940000e-01	1.236500e+00	-11.0645	2.172000e-01	4.625000e-01	8.366000e-01	1.753110e+01
ebitda_bev_x	2406313.0	5.730000e-02	1.310800e+00	-38.6063	9.750000e-02	1.837000e-01	2.972000e-01	3.290900e+00
ebit_bev_x	2406990.0	-2.510000e-02	1.386000e+00	-41.0563	5.220000e-02	1.282000e-01	2.282000e-01	2.8000000e+00
fi_bev_x	2116451.0	-8.600000e-02	1.345800e+00	-38.5103	4.190000e-02	9.910000e-02	1.608000e-01	2.274200e+00
cop_bev_x	2188818.0	3.139000e-01	8.344000e-01	-8.9448	8.920000e-02	2.259000e-01	4.111000e-01	$4.111000e\text{-}01 \big 1.607970e\text{+}01 \big $

	count	mean	std	min	25%	20%	12%	max
gp-ppen_x	2466653.0	2.766900e+00 6.510900e+00	6.510900e+00	-130.5385	4.559000e-01	1.518900e+00	1.518900e + 00 3.353000e + 00 1.035052e + 02	1.035052e+02
ebitda_ppen_x	2468488.0	-1.134000e-01	1.280070e+01	-558.0000	1.689000e-01	4.726000e-01	4.726000e-01 1.116300e+00 3.389320e+01	3.389320e+01
fcf_ppen_x	2270795.0	-8.658000e-01	1.104610e + 01	-423.4211	-3.778000e-01	-1.180000e-02	3.338000e-01	3.272670e+01
fincf_at_x	2181057.0	6.050000e-02	2.270000e-01	-0.9085	-4.100000e-02	1.800000e-03	8.120000e-02	8.120000e-02 1.643700e+00
netis_at_x	2180970.0	2.900000e-02	2.576000e-01	-1.3681	-4.860000e-02	0.0000000e+00	5.940000e-02	1.592800e+00
eqnetis_at_x	2181226.0	5.680000e-02	1.918000e-01	-0.3507	-8.000000e-04	6.000000e-04	1.520000e-02	1.488800e+00
eqis_at_x	2142004.0	7.050000e-02	1.912000e-01	-0.1034	0.0000000+00	3.200000e-03	2.280000e-02	1.535600e+00
dbnetis_at_x	2487875.0	-2.120000e-02	1.573000e-01	-1.3624	-3.980000e-02	-8.000000e-04	2.270000e-02	6.456000e-01
dltnetis_at_x	2487184.0	-2.430000e-02	1.364000e-01	-1.2268	-3.180000e-02	-2.200000e-03	1.200000e-03	5.184000e-01
dstnetis_at_x	2428021.0	3.5000000e-03	6.050000e-02	-0.4789	-5.100000e-03	0.0000000e+00	1.130000e-02	4.836000e-01
eqnpo_at_x	2177364.0	-4.470000e-02	1.949000e-01	-1.4673	-1.110000e-02	8.000000e-04	2.020000e-02	4.462000e-01
eqbb_at_x	2059717.0	1.250000e-02	3.500000e-02	-0.0026	0.00000000+00	0.00000000+00	5.300000e-03	4.018000e-01
div_at_x	2500964.0	1.160000e-02	2.170000e-02	0.0000	0.0000000e+00	1.900000e-03	1.660000e-02	3.183000e-01
oaccruals_at_x	2261617.0	-1.580000e-02	1.522000e-01	-2.2637	-7.200000e-02	-1.830000e-02	4.760000e-02	6.719000e-01
oaccruals_ni_x	2260635.0	$ \ -5.853000 e - 01 \ 6.180500 e + 00$	6.180500e+00	-71.4418	-1.208700e+00	-2.712000e-01	6.967000e-01	8.515790e+01
taccruals_at_x	2240180.0	-3.100000e-02	2.045000e-01	-2.4802	-9.100000e-02	-1.180000e-02	4.930000e-02	1.294200e+00
taccruals_ni_x	2238904.0	$2238904.0 \left -1.448100e + 00 \right 8.683400e + 00$	8.683400e+00	-131.5096	-1.516600e+00	-1.946000e-01	7.622000e-01	6.728570e + 01
noa_at_x	2142866.0	6.816000e-01	4.649000e-01	-1.1515	4.896000e-01	6.884000e-01	8.418000e-01	1.038840e+01
be_bev_x	2368048.0	1.343100e+00 2.666700e+00	2.666700e+00	0.0326	5.543000e-01	8.086000e-01	1.190400e+00	6.053070e + 01
debt_bev_x	2416506.0	4.732000e-01	6.162000e-01	0.0000	1.399000e-01	3.804000e-01	6.012000e-01	1.276120e+01
cash_bev_x	2397575.0	8.357000e-01	8.357000e-01 3.110100e+00	0.0000	3.800000e-02	1.245000e-01	4.276000e-01	8.007360e + 01
pstk_bev_x	2418755.0	2.720000e-02	1.704000e-01	0.0000	0.00000000+00	0.0000000e+00	0.0000000e+00	7.089400e+00
debtlt_bev_x	2412477.0	3.446000e-01	4.482000e-01	0.0000	5.390000e-02	2.671000e-01	4.815000e-01	9.026500e+00
debtst_bev_x	2403343.0	1.233000e-01	2.903000e-01	0.0000	3.200000e-03	3.390000e-02	1.172000e-01	5.633000e+00
int_debt_x	1959042.0	1.258000e-01	3.153000e-01	0.0000	5.310000e-02	7.610000e-02	1.063000e-01	7.750000e+00
int_debtlt_x	1874541.0	3.393000e-01	1.552500e+00	0.0000	6.360000e-02	9.400000e-02	1.485000e-01	4.145000e+01
ebitda_debt_x	2242375.0	2.161600e+00 2.312980e+01	2.312980e+01	-362.2105	1.666000e-01	3.959000e-01	9.501000e-01	5.562212e + 02
$profit_cl_x$	2270271.0	4.298000e-01	1.566600e+00	-11.9038	2.114000e-01	5.648000e-01	$ 1.016300\mathrm{e}{+00} 6.155300\mathrm{e}{+00}$	6.155300e+00
ocf_cl_x	2269486.0	5.390000e-02 1	1.456200e+00	-14.9568	-1.363000e-01	2.183000e-01	5.993000e-01	5.993000e-01 5.976400e+00

	count	mean	std	min		%02	75%	max
ocf_debt_x	2189764.0	1.253200e+00	1.968000e+01	-264.1167	-7.590000e-02	1.564000e-01	5.185000e-01	5.185000e-01 4.307215e+02
cash_lt_x	2487462.0	7.781000e-01	2.113200e+00	0.0000	4.150000e-02	1.312000e-01	5.084000e-01	2.990910e+01
inv_act_x	2124755.0	2.719000e-01	2.276000e-01	0.0000	4.860000e-02	2.538000e-01	4.448000e-01	9.113000e-01
rec_act_x	2130411.0	3.499000e-01	2.071000e-01	0.0000	1.990000e-01	3.479000e-01	4.754000e-01	9.455000e-01
debtst_debt_x	2235158.0	2.916000e-01	3.181000e-01	0.0000	3.900000e-02	1.578000e-01	4.582000e-01	1.000000e+00
cl_tx	2271050.0	5.408000e-01	2.822000e-01	0.0172	3.033000e-01	5.188000e-01	7.861000e-01	1.000000e+00
debtlt_debt_x	2251637.0	7.215000e-01	3.158000e-01	0.0000	5.637000e-01	8.571000e-01	9.724000e-01	1.0000000+00
lt_ppen_x	2467297.0	1.413180e+01	4.095230e+01	0.0809	1.032300e+00	2.019600e+00	5.768200e+00	7.630447e+02
debtlt_be_x	2439883.0	7.140000e-01	1.464700e+00	0.0000	3.360000e-02	3.025000e-01	7.618000e-01	2.225160e+01
opex_at_x	2503218.0	9.413000e-01	8.196000e-01	0.0029	3.295000e-01	7.872000e-01	1.304500e+00	7.158500e+00
nwc_at_x	2253296.0	2.724000e-01	2.457000e-01	-0.7924	8.520000e-02	2.536000e-01	4.349000e-01	9.547000e-01
debt_at_x	2514980.0	2.331000e-01	2.095000e-01	0.0000	5.090000e-02	1.957000e-01	3.591000e-01	1.428700e+00
debt_be_x	2444508.0	9.825000e-01	1.972300e+00	0.0000	9.520000e-02	4.426000e-01	1.023800e+00	3.440000e+01
ebit_int_x	2038745.0	1.266250e+01	1.784445e+02	-3702.0000	1.253300e+00	4.003000e+00 1.124330e+01	1.124330e+01	3.302250e+03
inv_days_x	2394275.0	8.869850e+01	1.683021e+02	0.0000	9.009300e+00	5.392190e+01	5.392190e+01 1.091676e+02	3.574195e+03
rec_days_x	2403668.0	3.602296e + 02	9.967740e+02	0.0000	3.863530e+01	5.827670e+01	8.822010e+01	7.354934e+03
ap_days_x	2314657.0	1.459695e+03	7.489965e+03	0.7812	2.587680e+01	4.209780e+01	7.865320e+01	1.412089e+05
cash_conversion_x	1836443.0	1.256743e+02	2.122532e+02	0.0000	4.172550e+01	8.193360e+01	1.398610e+02	3.521431e+03
cash_cl_x	2262167.0	1.419800e+00	3.231200e+00	0.0000	1.124000e-01	3.726000e-01	1.177400e+00 3.650000e+01	3.650000e+01
caliq_cl_x	2241081.0	$2241081.0 \mid 2.487700e+00 \mid$	3.827100e+00	0.0581	9.004000e-01	1.378900e+00	1.378900e+00 2.376600e+00 4.066670e+01	4.066670e+01
ca_cl_x	2252774.0	3.162200e+00	3.912700e+00	0.0824	1.372500e+00	2.102000e+00	2.102000e+00 3.307100e+00 4.119530e+01	4.119530e+01
inv_turnover_x	1990611.0	1.861590e+01	4.951140e+01	0.0438	2.956600e+00	5.130900e+00	1.205000e+01	7.307939e+02
at_turnover_x	2482416.0		9.318000e-01	0.0000	3.768000e-01	9.269000e-01	1.525100e+00	9.298300e+00
rec_turnover_x	2400338.0		2.636800e+01	0.0000	4.039600e+00	6.187900e+00	9.236800e+00	2.787135e+02
ap_turnover_x	2229997.0		1.238900e+01	-0.1258	4.826800e+00	8.918500e+00	8.918500e+00 1.434510e+01	1.336129e+02
sale_bev_x	2408388.0	2.269200e+00	2.923100e+00	0.0000	7.623000e-01	1.580300e+00	1.580300e + 00 2.598800e + 00 3.887110e + 01	3.887110e+01
sale_be_x	2437063.0	2.732600e+00	3.718300e+00	0.0000	9.001000e-01	1.758000e+00		5.438940e+01
div_ni_x	1963756.0	3.126000e-01		0.0000	0.0000000e+00	1.650000e-01	4.135000e-01	$4.135000e\text{-}01 \left 1.293670e\text{+}01 \right $
sale_nwc_x	2017664.0	9.746900e+00	2.267620e+01	0.0000	2.0669000e+00	3.971600e+00	3.971600e + 00 7.750900e + 00 3.110241e + 02	3.110241e+02

	count	mean	std	min	25%	20%	75%	max
tax_pi_x	1999061.0	3.279000e-01	3.117000e-01	-7.2981	2.705000e-01	3.654000e-01	4.329000e-01	5.548900e+00
cash_at_x	2496082.0	1.581000e-01	2.035000e-01	0.0000	2.580000e-02	7.260000e-02	2.026000e-01	9.799000e-01
ni_emp_x	2332173.0	$2332173.0 \left -1.044570e + 01 \right $	1.898294e+02	-3810.3810	4.055000e-01	4.200600e+00	1.703640e+01	1.438498e+03
sale_emp_x	2328826.0	2.691786e + 02	5.003031e+02	0.0000	6.301400e+01	1.411000e+02	2.763478e + 02	7.782523e+03
$sale_emp_gr1_x$	2120715.0	1.123000e-01	4.553000e-01	-0.9563	-3.330000e-02	5.300000e-02	1.513000e-01	7.027000e+00
emp_gr1_x	2048454.0	7.670000e-02	2.504000e-01	-1.3333	-3.060000e-02	4.520000e-02	1.538000e-01	1.483100e+00
ni_inc8q_x	1837805.0	3.116800e+00	3.262400e+00	0.0000	0.0000000+00	2.0000000e+00	7.000000e+00 8.000000e+00	8.0000000e+00
noa_gr1a_x	2130139.0	1.277000e-01	4.002000e-01	-0.7366	-1.750000e-02	4.940000e-02	1.574000e-01	1.075230e+01
ppeinv_gr1a_x	2130674.0	1.104000e-01	2.282000e-01	-0.5663	9.400000e-03	5.870000e-02	1.436000e-01	3.078700e+00
$lnoa_gr1a_x$	2042945.0	3.180000e-02	9.170000e-02	-0.5778	-3.800000e-03	1.370000e-02	4.740000e-02	7.544000e-01
capx_gr2_x	1996106.0	1.219100e+00	4.305300e+00	-1.4277	-2.477000e-01	2.272000e-01	1.043000e+00	7.697220e+01
saleq_gr1_x	2256822.0	2.428000e-01	8.315000e-01	-1.0000	-1.270000e-02	9.890000e-02	2.606000e-01	1.574840e+01
niq-be-x	2153966.0	5.000000e-04	1.393000e-01	-2.0216	1.600000e-03	2.420000e-02	4.290000e-02	6.993000e-01
niq_at_x	2218680.0	-2.200000e-03	6.080000e-02	-0.6672	0.0000000+000	8.200000e-03	2.060000e-02	1.818000e-01
niq-be-chg1_x	1961181.0	-7.700000e-03	1.339000e-01	-2.0038	-1.650000e-02	-6.000000e-04	1.090000e-02	1.227600e+00
niq-at-chg1_x	2044996.0	3.000000e-04	5.400000e-02	-0.4547	-7.100000e-03	-0.0000000e+00	5.6000000e-03	8.413000e-01
dsale_dinv_x	1796036.0	-4.380000e-02	8.780000e-01	-19.4778	-1.460000e-01	2.150000e-02	1.949000e-01	5.598300e+00
dsale_drec_x	2136436.0	-3.080000e-02	6.202000e-01	-7.3996	-1.418000e-01	1.500000e-03	1.418000e-01	1.418000e-01 7.637700e+00
dgp_dsale_x	2120443.0	2.720000e-02	5.405000e-01	-5.9700	-7.530000e-02	2.300000e-03	8.380000e-02	1.201120e+01
dsale_dsga_x	1827645.0	2.310000e-02	3.643000e-01	-2.2251	-8.920000e-02	-1.000000e-04	9.360000e-02	9.360000e-02 6.963700e+00
saleq_su_x	1944544.0	1.618000e-01	1.699500e+00	-16.0960	-8.666000e-01	1.532000e-01	1.125000e+00	.125000e+00 3.358810e+01
niq_su_x	1972831.0	-1.123000e-01	1.940400e+00	-50.8463	-7.565000e-01	5.100000e-03	7.529000e-01	2.019490e+01
capex_abn_x	1806456.0	1.173000e-01	9.626000e-01	-1.1469	-3.685000e-01	-6.920000e-02	2.932000e-01	1.196350e+01
op_atll_x	2415570.0	1.320000e-01	2.472000e-01	-6.9463	4.860000e-02	1.355000e-01	2.227000e-01	1.125400e+00
gp-atll_x	2413733.0	3.639000e-01	3.763000e-01	-1.9036	1.162000e-01	3.032000e-01	5.356000e-01	2.788000e+00
ope_bell_x	2010286.0	2.202000e-01	6.575000e-01	-13.6285	1.063000e-01	2.425000e-01	3.880000e-01	4.617600e+00
cop_atl1_x	2237311.0	1.409000e-01	2.863000e-01	-3.8344	4.500000e-02	1.505000e-01	2.563000e-01	
pi_nix_x	1959639.0	1.615000e+00	6.861000e-01	0.1059	1.340700e+00	1.572900e+00	1.777900e+00	\vdash
ocf_at_x	2449158.0	1.150000e-02	1.872000e-01	-1.8184	-2.140000e-02	4.090000e-02	1.033000e-01	5.979000e-01

std
1.575000e-01
1.627000e-01
1.149800e+01
4.732000e-01
1.916000e-01 0.0025
1.037400e+00
8.005000e-01
1.728500e+00 0.0000
3.035200e+00 -9.3872
9.357000e+00 -37.3359
5.258077e+03 0.0982
5.190800e+01 $ -1723.5716$
1.939000e-01 -1.1201
5.543000e-01 -7.5143
.917000e-01 -1.6595
.825000e-01 -0.9610
.593000e-01 -24.3597
6.110000e-01 -3.9640
1.054000e-01 0.0003
4.953100e+00 0.0086
7.141000e-01 0.0050
1.647800e+00
1.480500e+00 -3.4965
5.843000e-01
3.088100e+00
6.845000e-01
2.707000e-01
2.455000e-01 -7.4186

	count	mean	std	min	25%	%02	15%	max
ope_me_x	2183835.0	1.085000e-01	2.516000e-01	-8.0248	3.920000e-02	1.084000e-01	1.911000e-01	3.793500e+00
ni_me_x	2517298.0	1.200000e-03	3.459000e-01	-18.9294	5.500000e-03	4.900000e-02	8.530000e-02	9.917000e-01
nix_me_x	2517298.0	-1.200000e-03	3.693000e-01	-20.3694	4.100000e-03	4.920000e-02	8.650000e-02	1.036200e+00
cop_me_x	2259562.0	2.183000e-01	5.014000e-01	-3.5452	4.550000e-02	1.406000e-01	2.768000e-01	2.124680e+01
ocf_me_x	2450553.0	4.280000e-02	2.747000e-01	-5.6691	-1.830000e-02	5.360000e-02	1.205000e-01	5.711200e+00
fcf_me_x	2303306.0	-7.030000e-02	3.536000e-01	-8.5448	-1.065000e-01	-2.600000e-03	5.530000e-02	4.202300e+00
div_me_x	2501593.0	1.780000e-02	2.950000e-02	0.0000	0.00000000+000	3.900000e-03	2.660000e-02	1.049700e+00
eqbb_me_x	2059868.0	1.380000e-02	3.780000e-02	-0.0037	0.0000000+00	0.0000000e+00	7.800000e-03	8.704000e-01
eqis-me-x	2142182.0	4.550000e-02	1.388000e-01	-0.1339	1.000000e-04	3.5000000e-03	1.830000e-02	5.839400e+00
eqpo_me_x	2058263.0	3.150000e-02	5.660000e-02	-0.0013	0.00000000+000	1.120000e-02	4.150000e-02 1.725500e+00	1.725500e+00
eqnpo_me_x	2177501.0	-1.430000e-02	1.450000e-01	-6.1142	-8.100000e-03	1.200000e-03	3.130000e-02	1.442900e+00
eqnetis_me_x	2181408.0	3.130000e-02	1.401000e-01	-0.6866	-1.400000e-03	7.000000e-04	1.260000e-02	5.679700e+00
at_mev_x	2480516.0	1.759600e+00	3.280300e+00	0.0085	5.638000e-01	1.008000e+00	1.587100e + 00 6.916660e + 01	6.916660e+01
bev_mev_x	2404633.0	6.919000e-01	5.487000e-01	0.0009	3.194000e-01	6.308000e-01	9.482000e-01	1.692550e+01
ppen_mev_x	2459710.0	3.322000e-01	3.872000e-01	0.0000	5.950000e-02	1.893000e-01	4.753000e-01	6.654400e+00
be_mev_x	2410201.0	6.153000e-01	8.336000e-01	0.0050	2.513000e-01	4.357000e-01	7.057000e-01	2.914710e+01
cash_mev_x	2460357.0	2.333000e-01	6.6110000e-01	0.0000	2.350000e-02	6.940000e-02	1.825000e-01	1.486960e+01
sale_mev_x	2472091.0	1.265200e+00	1.765600e+00	0.0000	3.146000e-01	7.343000e-01	1.550900e+00	3.775600e+01
gp_mev_x	2467238.0	3.453000e-01	4.647000e-01	-2.4081	1.209000e-01	2.305000e-01	4.284000e-01	1.314000e+01
ebitda_mev_x	2469299.0	1.012000e-01	2.101000e-01	-5.5869	5.090000e-02	1.060000e-01	1.669000e-01	2.711700e+00
ebit_mev_x	2470075.0	6.010000e-02	2.226000e-01	-6.8743	2.670000e-02	7.470000e-02	1.222000e-01	2.601300e+00
cop_mev_x	2243652.0	1.516000e-01	2.798000e-01	-2.3844	4.200000e-02	1.203000e-01	2.126000e-01	8.747500e+00
ocf_mev_x	2431339.0	3.150000e-02	1.968000e-01	-4.7377	-1.650000e-02	4.340000e-02	9.350000e-02	2.334400e+00
fcf_mev_x	2286863.0	-3.800000e-02	2.261000e-01	-6.0410	-8.630000e-02	-2.300000e-03	4.670000e-02	1.728000e+00
debt_mev_x	2480615.0	3.008000e-01	3.502000e-01	0.0000	4.020000e-02	2.106000e-01	4.607000e-01	7.224300e+00
pstk_mev_x	2479267.0	1.480000e-02	5.860000e-02	0.0000	0.0000000e+00	0.0000000e+00	0.0000000e+00 1.220500e+00	1.220500e+00
debtlt_mev_x	2476104.0	2.224000e-01	2.443000e-01	0.0000	1.400000e-02	1.446000e-01	3.542000e-01	2.411300e+00
$debtst_mev_x$	2461067.0	8.090000e-02	2.125000e-01	0.0000	9.000000e-04	1.690000e-02	7.010000e-02	5.292900e+00
dltnetis_mev_x	2453443.0	-3.150000e-02	1.853000e-01	-3.5613	-3.440000e-02	-1.900000e-03	1.200000e-03	6.324000e-01

	count	mean	std	min	25%	%02	15%	max
dstnetis_mev_x	2393968.0	4.100000e-03	9.390000e-02	-1.0163	-4.800000e-03	0.00000000+000	1.110000e-02	1.122900e+00
dbnetis_mev_x	2454176.0	-2.880000e-02	2.223000e-01	-4.4848	-4.210000e-02	-6.000000e-04	2.280000e-02	2.280000e-02 1.188800e+00
netis_mev_x	2164671.0	-8.300000e-03	2.729000e-01	-4.6395	-5.040000e-02	0.00000000+00	5.030000e-02	5.358400e+00
fincf_mev_x	2164802.0	3.700000e-02	2.405000e-01	-2.3006	-4.040000e-02	1.300000e-03	7.090000e-02	6.822000e+00
aliq_mat_x	2036506.0	5.016000e-01	2.661000e-01	0.0270	3.052000e-01	4.793000e-01	6.504000e-01	3.973200e+00
eq_dur_x	2193667.0	1.598720e+01	5.630900e+00	0.2861	1.413720e+01	1.612420e+01	1.764670e+01	3.430355e+02
$beta_60m_x$	2090801.0	1.153800e+00	6.856000e-01	-1.7467	6.897000e-01	1.081600e+00	1.528500e+00	4.912400e+00
ivol_capm_60m_x	2090801.0	1.172000e-01	6.560000e-02	0.0288	7.050000e-02	1.002000e-01	1.454000e-01	5.392000e-01
resff3_12_1_x	2274040.0	-2.210000e-02	2.736000e-01	-1.1550	-1.908000e-01	-8.900000e-03	1.610000e-01	7.899000e-01
$resff3_6_1x$	2273172.0	-5.420000e-02	5.396000e-01	-2.9537	-3.435000e-01	-2.040000e-02	2.734000e-01	1.925800e+00
mispricing_mgmt_x	2414716.0	4.896000e-01	1.856000e-01	0.0147	3.610000e-01	5.047000e-01	6.284000e-01	9.427000e-01
mispricing_perf_x	2649116.0	5.208000e-01	2.065000e-01	0.0099	3.773000e-01	5.270000e-01	6.749000e-01	9.881000e-01
$zero_trades_21d_x$	2568596.0	9.102000e-01	9.102000e-01 2.670500e+00	0.0000	1.800000e-03	3.700000e-03	7.200000e-03	2.100980e+01
$dolvol_126d_x$	2527407.0	1.272436e+07	5.041472e+07	36.1000	6.023594e+04	3.756701e+05	3.493927e+06	1.038495e + 09
$dolvol_var_126d_x$	2527340.0	1.275800e+00	7.751000e-01	0.2622	7.587000e-01	1.088500e+00 1.545100e+00	1.545100e+00	8.289100e+00
$turnover_126d_x$	2527415.0	4.300000e-03	6.8000000e-03	0.0000	9.000000e-04	2.200000e-03	5.300000e-03	2.857000e-01
$turnover_var_126d_x$	2527348.0	1.251900e+00	7.609000e-01	0.2796	7.459000e-01	1.058700e+00	1.509700e+00	7.678300e+00
$zero_trades_126d_x$	2527415.0	9.170000e-01	9.170000e-01 2.511100e+00	0.0000	1.900000e-03	4.000000e-03	1.771000e-01	1.949730e+01
$zero_trades_252d_x$	2472485.0	9.236000e-01	9.236000e-01 2.470600e+00	0.0001	2.000000e-03	4.300000e-03	2.625000e-01	1.910030e+01
$bidaskhl_21d_x$	2474735.0	1.470000e-02	1.810000e-02	0.0011	5.600000e-03	9.400000e-03	1.710000e-02	5.318000e-01
$rolh1_21d_x$	2474735.0	2.130000e-02	1.570000e-02	0.0000	1.100000e-02	1.720000e-02	2.680000e-02	1.854000e-01
beta_21d_x	2469080.0	8.736000e-01	1.205700e+00	-11.1429	2.238000e-01	8.042000e-01	1.458000e+00	1.276490e+01
$ivol_capm_21d_x$	2469080.0	2.710000e-02	1.960000e-02	0.0018	1.400000e-02	2.160000e-02	3.380000e-02	2.415000e-01
iskew_capm_21d_x	2469046.0	2.407000e-01	8.745000e-01	-3.5665	-2.542000e-01	2.053000e-01	7.097000e-01	3.715300e+00
coskew_21d_x	2469074.0	-1.530000e-02	3.111000e-01	-1.4678	-2.232000e-01	-2.070000e-02	1.886000e-01	1.347500e+00
beta_dimson_21d_x	2469080.0	9.503000e-01	1.950600e+00	-19.3713	4.290000e-02	8.515000e-01	1.798400e+00	2.341690e+01
ivol_ff3_21d_x	2469080.0	2.640000e-02	1.930000e-02	0.0018	1.360000e-02	2.100000e-02	3.300000e-02	2.340000e-01
iskew_ff3_21d_x	2469068.0	1.990000e-01	7.943000e-01	-3.1203	-2.632000e-01	1.696000e-01	6.344000e-01	3.455800e+00
ivol_hxz4_21d_x	2332649.0	2.680000e-02	1.960000e-02	0.0018	1.370000e-02	2.130000e-02	3.350000e-02	2.397000e-01

	count	mean	std	min	25%	20%	12%	max
iskew_hxz4_21d_x	2332643.0	1.777000e-01	7.585000e-01	-3.0805	-2.681000e-01	1.513000e-01	$1.513000 \mathrm{e}\text{-}01 \ 6.005000 \mathrm{e}\text{-}01 3.275600 \mathrm{e}\text{+}00$	3.275600e+00
rmax5_21d_x	2469033.0	3.860000e-02	2.910000e-02	0.0022	1.960000e-02	3.050000e-02		4.810000e-02 $3.544000e-01$
rmax1_21d_x	2469033.0	6.730000e-02	5.830000e-02	0.0035	3.110000e-02	5.000000e-02	8.280000e-02	8.996000e-01
rvol_21d_x	2469080.0	2.970000e-02	2.060000e-02	0.0018	1.590000e-02	2.400000e-02	3.690000e-02	2.515000e-01
rskew_21d_x	2469038.0	2.439000e-01	8.740000e-01	-3.5810	-2.529000e-01	2.077000e-01	7.136000e-01	7.136000e-01 3.808400e+00
ami_126d_x	2427976.0	2427976.0 $ 2.294900e+00$	1.277990e+01	0.0000	6.300000e-03	8.310000e-02	7.621000e-01	7.621000e-01 $ 7.242321e+02 $
$beta_252d_x$	2434576.0	8.972000e-01	6.011000e-01	-1.8325	4.682000e-01	8.481000e-01	8.481000e-01 1.259200e+00 4.013900e+00	4.013900e+00
$ ivol_capm_252d_x $	2434576.0	2.910000e-02	1.710000e-02	0.0050	1.700000e-02	2.480000e-02	3.650000e-02 1.684000e-01	1.684000e-01
betadown_252d_x	2406390.0	1.001300e+00	7.817000e-01	-3.9821	5.127000e-01	9.352000e-01	9.352000e-01 1.414000e+00 5.699200e+00	5.699200e+00
$prc_highprc_252d_x$	2434268.0	7.724000e-01	1.997000e-01	0.0167	6.6100000e-01	8.272000e-01		9.318000e-01 1.000000e+00
rol_252d_x	2434576.0	3.110000e-02	1.740000e-02	0.0052	1.870000e-02	2.680000e-02	3.870000e-02	3.870000e-02 $1.690000e-01$
corr_1260d_x	1904407.0	3.603000e-01	1.650000e-01	-0.0374	2.362000e-01	3.573000e-01	4.786000e-01	8.219000e-01
$betabab_1260d_x$	1893789.0	1.075600e+00	5.871000e-01	-0.3259	6.475000e-01	1.000500e+00	1.000500e+00 1.410600e+00 4.274900e+00	4.274900e+00
rmax5_rvol_21d_x	2343331.0	1.232900e+00	5.049000e-01	0.1125	8.720000e-01	1.159900e+00	1.159900e+00 1.512600e+00 4.328700e+00	4.328700e+00
age_x	2739928.0	2739928.0 $2.184690e+02$	1.888040e+02	1.0000	7.900000e+01	1.590000e+02	1.590000e + 02 3.000000e + 02 1.115000e + 03	1.115000e+03
qmj_x	1825615.0	8.990000e-02	9.763000e-01	-1.7027	-7.318000e-01	1.204000e-01		9.350000e-01 1.701100e+00
qmj_prof_x	2502382.0	9.110000e-02	9.846000e-01	-1.7036	-7.339000e-01	1.300000e-01	9.456000e-01	9.456000e-01 1.698800e+00
$ qmj_growth_x $	1825622.0	3.610000e-02	9.739000e-01	-1.7018	-7.911000e-01	4.900000e-02	8.716000e-01	8.716000e-01 1.702100e+00
qmj_safety_x	2579701.0	8.730000e-02	9.713000e-01	-1.7012	-7.189000e-01	1.215000e-01	9.239000e-01	9.239000e-01 1.708800e+00

Table 4: Complete Global Factor Dataset Summary Statistics

	count	mean	std	min	25%	20%	12%	max
beta_60m_x	532218.0	532218.0 $1.171000e+00$	4.530000e-01	-0.350	0.886	1.149	1.393000e+00 3.3890	3.389000e+00
index	532218.0	8.806473e+05 $4.668442e+05$	4.668442e+05	651.000	528652.250	821568.500	1.220481e+06 2.733106e+06	90+ - 90
	532218.0	1.976878e + 05	8.005020e+02	196101.000	197103.000	197711.000	1.984030e + 05 1.9891	1.989120e + 05
permno	532218.0	3.638201e+04	1.839758e + 04	10006.000	21573.000	32302.000	4.972900e+04 9.3220	9.322000e+04
permco	532218.0	1.740521e+04 8.516931e+03	8.516931e+03	4.000	8611.000	21173.000	2.289400e+04 5.6285	5.628500e+04
crsp_shrcd	532218.0	1.066100e + 01	5.230000e-01	10.000	10.000	11.000	1.100000e+01 1.2000	1.200000e+01
crsp_exchcd	532218.0	1.509000e+00	8.170000e-01	1.000	1.000	1.000	2.000000e+00 3.0000	3.0000000e+00
	532218.0	4.147325e+03	1.685914e+03	100.000	2899.000	3721.000	5.311000e+03 9.5110	9.511000e+03
	532218.0	2.874800e+01	1.327400e+01	1.000	18.000	31.000	4.100000e+01 4.9000	4.900000e+01
adjfct	532218.0	6.901000e+00	2.505700e+01	0.000	1.000	2.000	4.0000000e+00 1.2150	1.215000e+03
shares	532218.0	1.992800e+01	4.235100e+01	0.212	4.200	8.553	1.963800e+01 1.4263	1.426344e+03
	532218.0	6.983350e + 02	2.292746e+03	17.030	94.162	196.351	5.360830e + 02 1.0202	1.020223e+05
me_company	532218.0		2.338138e+03	17.030	94.471	197.355	5.383670e + 02 1.0202	1.020223e+05
	532218.0	3.142100e+01	5.300700e+01	0.422	17.375	26.000	3.775000e+01 8.6750	8.675000e+03
prc_local	532218.0	3.142100e+01	5.300700e+01	0.422	17.375	26.000	3.775000e+01 8.6750	8.675000e+03
dolvol	532218.0	2.686159e + 07	1.110126e + 08	0.000	1586250.000	4175525.000	1.458036e+07 9.0278	9.027815e+09
	532218.0	1.700000e-02	1.090000e-01	-0.761	-0.043	0.008	6.800000e-02 3.1850	3.185000e+00
ret_local	532218.0	1.700000e-02	1.090000e-01	-0.761	-0.043	0.008	6.800000e-02 3.1850	3.185000e+00
ret_exc	532218.0	1.100000e-02	1.090000e-01	-0.771	-0.049	0.003	6.200000e-02 3.1800	3.180000e+00
ret_lag_dif	532218.0	1.0000000e+00 0.000000e+00	0.0000000+00	1.000	1.000	1.000	1.000000e+00 1.0000	1.0000000e+00
ret_exc_lead1m	532218.0	5.000000e-03	1.060000e-01	-1.005	-0.052	0.000	5.800000e-02 2.4720	2.472000e+00
niq_su_x	532218.0	3.200000e-02	1.438000e+00	-15.807	-0.293	0.033	4.670000e-01 1.6424	1.642400e+01
$ret_6_1_x$	532218.0	8.100000e-02	2.470000e-01	-0.811	-0.057	0.039	1.860000e-01 2.8110	2.811000e+00
ret_12_1x	532218.0	1.810000e-01	4.100000e-01	-0.859	-0.043	0.095	3.260000e-01 6.1880	6.188000e+00
saleq_su_x	532218.0	2.610000e-01	1.384000e+00	-11.313	-0.177	0.192	7.030000e-01 3.3588	3.358800e+01
tax_gr1a_x	532218.0	6.000000e-03	2.200000e-02	-0.153	-0.001	0.003	1.200000e-02 1.500	1.500000e-01
ni_inc8q_x	532218.0	3.814000e+00	2.800000e+00	0.000	3.000	3.000	7.000000e+00 8.0000	8.0000000e+00
$prc_highprc_252d_x$	532218.0	8.310000e-01	1.410000e-01	0.116	0.766	0.853	9.380000e-01 $ 1.0000$	1.000000e+00
$resff3_6_1_x$	532218.0	-4.200000e-02	5.010000e-01	-2.954	-0.282	-0.019	2.320000e-01 1.9260	1.926000e+00

	count	mean	std	min	25%	20%	75% max
$resff3_12_1x$	532218.0	-1.800000e-02	2.550000e-01	-1.155	-0.161	-0.007	1.370000e-01 $ 7.900000e-01$
be_me_x	532218.0	8.1100000e-01	5.520000e-01	0.012	0.448	0.724	1.022000e+00 1.009700e+01
debt_me_x	532218.0	7.170000e-01	1.271000e+00	0.000	0.123	0.332	7.470000e-01 2.510800e+01
at_me_x	532218.0	2.978000e+00	5.266000e+00	0.037	0.831	1.482	2.495000e+00 5.293000e+01
$ret_60_12_x$	532218.0	9.080000e-01	1.395000e+00	-0.939	0.286	0.550	1.046000e+00 2.063600e+01
ni_me_x	532218.0	8.300000e-02	9.300000e-02	-4.344	0.054	0.074	$1.150000e$ -01 $\mid 8.570000e$ -01
fcf_me_x	532218.0	-1.330000e-01	3.110000e-01	-6.788	-0.149	-0.070	-1.600000e-02 2.322000e+00
div12m_me_x	532218.0	3.200000e-02	2.500000e-02	0.000	0.013	0.028	4.600000e-02 1.750000e-01
eqpo_me_x	532218.0	3.300000e-02	3.900000e-02	-0.000	0.019	0.020	3.600000e-02 6.870000e-01
eqnpo_me_x	532218.0	1.300000e-02	6.100000e-02	-1.569	0.010	0.010	2.700000e-02 5.800000e-01
sale_gr3_x	532218.0	5.970000e-01	1.045000e+00	-0.983	0.250	0.396	6.210000e-01 2.835500e+01
sale_gr1_x	532218.0	1.680000e-01	2.990000e-01	-0.911	0.061	0.121	2.020000e-01 8.495000e+00
ebitda_mev_x	532218.0	1.650000e-01	1.360000e-01	-2.840	0.109	0.143	1.960000e-01 2.712000e+00
sale_me_x	532218.0	2.020000e+00	2.490000e+00	0.000	0.786	1.518	2.269000e+00 6.705800e+01
ocf_me_x	532218.0	1.500000e-02	2.220000e-01	-4.015	-0.020	0.020	8.100000e-02 2.748000e+00
intrinsic_value_x	532218.0	4.197170e+02	8.801200e+02	0.099	65.828	113.423	3.517070e+02 9.626163e+03
bev_mev_x	532218.0	7.700000e-01	4.030000e-01	0.009	0.505	0.773	9.770000e-01 $ 5.615000e$ +00
netdebt_me_x	532218.0	4.110000e-01	1.070000e+00	-3.343	0.016	0.201	5.340000e-01 2.390300e+01
eq_dur_x	532218.0	1.460200e+01	2.748000e+00	0.286	13.409	14.948	1.632600e+01 3.766200e+01
capex_abn_x	532218.0	3.600000e-02	5.400000e-01	-1.139	-0.142	-0.061	9.300000e-02 9.084000e+00
at_gr1_x	532218.0	1.790000e-01	3.130000e-01	-0.588	0.058	0.108	1.900000e-01 6.247000e+00
ppeinv_gr1a_x	532218.0	1.200000e-01	1.630000e-01	-0.566	0.058	0.088	1.370000e-01 2.396000e+00
noa_at_x	532218.0	7.700000e-01	2.170000e-01	-0.206	0.701	0.758	8.310000e-01 3.512000e+00
noa_gr1a_x	532218.0	1.000000e-01	1.790000e-01	-0.585	0.036	0.068	1.120000e-01 3.209000e+00
lnoa_gr1a_x	532218.0	2.800000e-02	4.300000e-02	-0.345	0.014	0.018	3.200000e-02 4.360000e-01
capx_gr1_x	532218.0	3.080000e-01	1.049000e+00	-1.263	-0.024	0.135	3.410000e-01 3.425000e+01
capx_gr2_x	532218.0	6.110000e-01	1.740000e+00	-1.324	0.069	0.278	5.800000e-01 $4.400000e+01$
capx_gr3_x	532218.0	8.810000e-01	2.349000e+00	-1.209	0.190	0.405	7.490000e-01 5.162800e+01
$chcsho_12m_x$	532218.0	3.300000e-02	1.100000e-01	-0.270	0.000	0.002	2.000000e-02 1.928000e+00

	count	mean	std	mim		20%	75%	max
<u>ਨੂੰ</u>	532218.0	7.000000e-03	9.200000e-02	-0.976	-0.000	0.021	4.500000e-02	4.740000e-01
<u>ත්</u>	532218.0		5.505000e+00	-1.000	0.099	0.318	7.220000e-01	1.709550e+02
<u> </u>	532218.0	1.780000e-01	4.590000e-01	-1.000	0.032	0.106	1.960000e-01	8.957000e+00
<u>ഹ</u>	532218.0	1.800000e-02	4.400000e-02	-0.353	0.000	0.008	2.700000e-02	2.980000e-01
<u> </u>	532218.0	6.800000e-02	9.900000e-02	-1.094	0.031	0.055	9.600000e-02	6.720000e-01
<u>5</u>	532218.0	4.600000e-02	8.300000e-02	-1.068	0.024	0.038	6.400000e-02	7.950000e-01
<u>v</u>	532218.0	2.300000e-02	5.700000e-02	-0.546	0.001	0.018	3.600000e-02	4.190000e-01
<u> </u>	532218.0	4.500000e-02	7.000000e-02	-0.618	0.015	0.036	6.000000e-02	4.920000e-01
<u>v</u>	532218.0	2.100000e-02	4.100000e-02	-0.372	0.005	0.017	3.200000e-02	3.650000e-01
<u> </u>	532218.0	4.600000e-02	7.400000e-02	-0.951	0.017	0.031	6.200000e-02	5.960000e-01
<u>ಸ</u>	532218.0	5.400000e-02	7.700000e-02	-0.965	0.022	0.037	7.200000e-02	5.770000e-01
<u>. 27</u>	532218.0	7.000000e-03	1.700000e-02	-0.180	0.001	0.003	1.000000e-02	1.700000e-01
<u>5</u>	532218.0	-2.100000e-02	9.400000e-02	-0.807	-0.049	-0.008	1.000000e-02	7.900000e-01
<u> </u>	532218.0	5.000000e-03	3.000000e-02	-0.496	0.000	0.000	0.00000000+000	3.480000e-01
<u>5.</u>	532218.0	2.700000e-02	8.600000e-02	-0.726	-0.005	0.008	4.900000e-02	6.180000e-01
<u>വ്</u>	532218.0	5.700000e-02	7.900000e-02	-0.977	0.022	0.045	7.500000e-02	7.450000e-01
വ്	532218.0	1.258000e+00	4.084000e+00	-32.889	0.431	0.794	1.390000e+00	8.515800e+01
<u>ഹ്</u>	532218.0	6.240000e-01	2.539000e+00	-84.700	0.425	0.640	8.670000e-01	4.606200e+01
<u>ഹ</u>	532218.0	2.000000e-02	9.400000e-02	-0.939	0.001	0.007	1.900000e-02	1.385000e+00
<u> </u>	532218.0	1.100000e-02	6.0000000e-02	-0.201	0.000	0.000	2.000000e-03	1.283000e+00
<u>5</u>	532218.0	1.700000e-02	8.400000e-02	-0.933	-0.008	0.000	4.000000e-02	6.460000e-01
<u> </u>	532218.0	3.200000e-02	3.600000e-02	-1.486	0.029	0.030	3.900000e-02	3.710000e-01
<u> </u>	532218.0	-1.000000e-03	3.500000e-02	-1.288	-0.001	-0.000	1.000000e-03	8.180000e-01
<u>v</u>	532218.0	1.500000e-02	1.600000e-02	-0.401	0.012	0.013	1.900000e-02	1.470000e-01
<u>5</u>	32218.0	532218.0 -0.000000e+00	1.300000e-02	-0.279	-0.000	-0.000	0.00000000+000	3.050000e-01
<u> </u>	532218.0	2.100000e-01	2.350000e-01	-7.292	0.118	0.153	2.450000e-01	2.122000e+00
<u>0</u>	532218.0	1.260000e-01	2.260000e-01	-17.327	0.077	0.103	1.770000e-01	5.410000e-01
5.	532218.0	2.096000e+00	1.939000e+00	0.000	1.145	1.796	2.305000e+00	2.035900e+01
<u>0</u>	532218.0	1.283000e+00	9.470000e-01	0.000	0.654	1.253	1.6000000e+00	1.600000e+00 9.298000e+00

	count	mean	std	min	25%	20%	75% r	max
gp_at_x	532218.0	3.330000e-01	2.380000e-01	-1.266	0.156	0.313	4.450000e-01 1.350000e+00	00+
gp_atll_x	532218.0	3.940000e-01	2.990000e-01	-1.629	0.179	0.360	5.150000e-01 2.081000e+00	00+
ope-be-x	532218.0	2.690000e-01	1.900000e-01	-5.541	0.213	0.245	3.270000e-01 1.636000e+00	00+
ope_bell_x	532218.0	3.220000e-01	2.540000e-01	-4.721	0.240	0.277	3.740000e-01 3.210000e+00	00+
op_at_x	532218.0	1.550000e-01	9.900000e-02	-1.233	0.107	0.140	2.010000e-01 5.210000e-01	e-01
op_atll_x	532218.0	1.830000e-01	1.300000e-01	-1.902	0.121	0.159	2.320000e-01 9.190000e-01	e-01
cop_at_x	532218.0	1.010000e-01	1.210000e-01	-1.188	0.065	0.098	1.490000e-01 8.660000e-01	e-01
cop_atll_x	532218.0	1.010000e-01	1.800000e-01	-3.468	0.074	0.107	1.620000e-01 8.260000e-01	e-01
f_score_x	532218.0	4.987000e+00	1.462000e+00	0.000	4.000	5.000	6.000000e+00 9.000000e+00	00+
O_SCOFe_X	532218.0	532218.0 -2.600000e+00	1.428000e+00	-8.827	-3.333	-2.108	-2.063000e+00 1.890400e+01	+01
Z_SCOTe_X	532218.0	4.292000e+00	3.95000000+000	-4.581	2.871	3.583	4.396000e+00 1.279700e+02	+05
pi_nix_x	532218.0	1.711000e+00	4.590000e-01	0.188	1.526	1.726	1.887000e+00 8.722000e+00	00+
at_be_x	532218.0	3.303000e+00	4.371000e+00	1.000	1.592	1.971	2.618000e+00 $4.812300e+01$	+01
saleq_gr1_x	532218.0	1.670000e-01	3.690000e-01	-0.986	0.066	0.117	1.900000e-01 1.273600e+01	+01
opex_at_x	532218.0	1.020000e+00	8.020000e-01	0.030	0.473	1.019	1.282000e+00 7.158000e+00	00+
emp_gr1_x	532218.0	6.200000e-02	1.610000e-01	-1.301	900.0	0.043	8.400000e-02 1.483000e+00	00+
age_x	532218.0	2.681630e + 02	1.779190e+02	1.000	129.000	234.000	3.870000e+02 7.680000e+02	+05
dsale_dinv_x	532218.0	-9.000000e-03	3.880000e-01	-5.274	-0.051	0.016	8.300000e-02 3.661000e+00	00+
dsale_drec_x	532218.0	-2.200000e-02	3.390000e-01	-6.314	-0.079	-0.002	6.700000e-02 3.915000e+00	00+
dgp_dsale_x	532218.0	2.300000e-02	3.430000e-01	-3.695	-0.045	0.001	5.100000e-02 1.201100e+01	+01
dsale_dsga_x	532218.0	-1.000000e-03	1.860000e-01	-2.225	-0.034	-0.006	2.500000e-02 3.436000e+00	00+
sale_emp_gr1_x	532218.0	8.900000e-02	1.970000e-01	-0.812	0.029	0.074	1.220000e-01 6.027000e+00	00+
tangibility_x	532218.0	6.880000e-01	1.050000e-01	0.137	0.650	0.685	7.300000e-01 1.389000e+00	00+
kz_index_x	532218.0	-3.152000e+00	7.543000e+00	-121.722	-3.120	-1.111	-5.090000e-01 1.529200e+01	+01
cash_at_x	532218.0	9.400000e-02	1.010000e-01	0.000	0.031	0.062	1.180000e-01 8.810000e-01	e-01
ni_ar1_x	532218.0	2.880000e-01	4.890000e-01	-2.015	0.080	0.243	4.840000e-01 3.490000e+00	00+
ni_ivol_x	532218.0	1.500000e-02	2.000000e-02	0.000	900.0	0.013	1.500000e-02 6.330000e-01	e-01
earnings_variability_x 532218.0	532218.0	3.330000e-01	2.800000e-01	0.024	0.181	0.269	3.420000e-01 3.436000e+00	00+
aliq_at_x	532218.0	7.560000e-01	2.310000e-01	0.240	999.0	0.727	7.750000e-01 5.901000e+00	00+

	count	mean	std	min	25%	20%	%52	max
aliq_mat_x	532218.0	5.760000e-01	2.150000e-01	0.041	0.477	0.586	6.460000e-01	3.973000e+00
seas_1_1an_x	532218.0	1.500000e-02	1.020000e-01	-0.591	-0.041	0.008	6.300000e-02	1.250000e+00
seas_1_1na_x	532218.0	1.600000e-02	2.700000e-02	-0.141	0.005	0.013	2.500000e-02	2.410000e-01
seas_2_5an_x	532218.0	1.500000e-02	4.400000e-02	-0.252	-0.002	0.012	2.900000e-02	5.730000e-01
ivol_ff3_21d_x	532218.0	1.800000e-02	9.000000e-03	0.004	0.012	0.016	2.100000e-02	1.330000e-01
ivol_capm_252d_x	532218.0	2.000000e-02	8.000000e-03	0.007	0.014	0.019	2.300000e-02	1.030000e-01
$-capm_21d_x$	532218.0	1.800000e-02	9.000000e-03	0.004	0.012	0.017	2.200000e-02	1.340000e-01
ivol_hxz4_21d_x	532218.0	1.900000e-02	8.000000e-03	0.005	0.013	0.020	2.000000e-02	1.290000e-01
rvol_21d_x	532218.0	2.000000e-02	1.100000e-02	0.004	0.013	0.019	2.400000e-02	1.560000e-01
$betabab_1260d_x$	532218.0	1.152000e+00	4.620000e-01	0.106	0.874	1.109	1.340000e+00	4.226000e+00
beta_dimson_21d_x	532218.0	9.880000e-01	1.424000e+00	-19.371	0.274	0.848	1.630000e+00	2.341700e+01
$turnover_126d_x$	532218.0	2.000000e-03	2.000000e-03	0.000	0.001	0.001	2.000000e-03	1.500000e-02
turnover_var_126d_x	532218.0	1.131000e+00	5.060000e-01	0.344	0.813	1.059	1.254000e+00	6.403000e+00
$dolvol_126d_x$	532218.0	1.040906e+06	2.747060e+06	175.564	85266.781	208631.680	6.878182e+05	2.997359e+07
$dolvol_var_126d_x$	532218.0	1.1600000e+00	5.200000e-01	0.363	0.834	1.090	1.288000e+00	6.326000e+00
ami_126d_x	532218.0	3.380000e-01	1.263000e+00	0.000	0.033	0.128	3.250000e-01	2.019330e+02
$zero_trades_21d_x$	532218.0	3.070000e-01	1.367000e+00	0.000	0.002	0.005	7.0000000e-03	2.009700e+01
$zero_trades_126d_x$	532218.0	3.050000e-01	1.240000e+00	0.000	0.003	0.006	8.0000000e-03	1.904100e+01
$zero_trades_252d_x$	532218.0	3.050000e-01	1.201000e+00	0.000	0.003	0.006	9.000000e-03	1.861000e+01
$rmax1_21d_x$	532218.0	4.600000e-02	3.100000e-02	0.006	0.026	0.040	5.500000e-02	4.290000e-01
$rskew_2ld_x$	532218.0	2.520000e-01	7.770000e-01	-2.611	-0.195	0.247	6.720000e-01	3.177000e+00
$iskew_capm_21d_x$	532218.0	2.350000e-01	7.530000e-01	-2.841	-0.202	0.227	6.370000e-01	3.099000e+00
$iskew_f3_21d_x$	532218.0	1.930000e-01	6.900000e-01	-2.111	-0.212	0.186	5.650000e-01	2.802000e+00
iskew_hxz4_21d_x	532218.0	1.690000e-01	6.030000e-01	-2.064	-0.115	0.165	4.170000e-01	2.6800000e+00
coskew_21d_x	532218.0	-1.700000e-02	3.000000e-01	-1.437	-0.206	-0.026	1.660000e-01	1.300000e+00
ret_1_0x	532218.0	1.600000e-02	1.040000e-01	-0.606	-0.041	0.008	6.500000e-02	1.167000e+00
$betadown_252d_x$	532218.0	1.050000e+00	6.690000e-01	-3.043	0.615	0.938	1.373000e+00	5.248000e+00
bidaskhl_21d_x	532218.0	7.000000e-03	5.000000e-03	0.001	0.004	0.006	8.0000000e-03	2.540000e-01
ret_3_1_x	532218.0	3.200000e-02	1.480000e-01	-0.717	-0.049	0.014	1.030000e-01	1.556000e+00

	count	mean	std	mim	25%	20%	%22	max
ret_9_1_x	532218.0	532218.0 1.310000e-01 3.340000e-01	3.340000e-01	-0.836	-0.054	0.066		2.600000e-01 $4.818000e+00$
ret_12_7x	532218.0		7.700000e-02 2.420000e-01	-0.779	-0.055	0.041	1.770000e-01	1.770000e-01 2.934000e+00
$corr_1260d_x$	532218.0	4.190000e-01	9.900000e-02	0.040	0.376	0.389	4.820000e-01	4.820000e-01 $7.180000e-01$
$rmax5_21d_x$	532218.0	532218.0 $2.700000e-02$ $1.500000e-02$	1.500000e-02	0.003	0.016	0.024	3.200000e-02	3.200000e-02 2.040000e-01
rmax5_rvol_21d_x	532218.0	532218.0 $1.246000e+00$ $4.630000e-01$	4.630000e-01	0.141	0.933	1.193	1.193 1.486000e+00 3.492000e+00	3.492000e+00
ni_be_x	532218.0	1.210000e-01	1.370000e-01	-7.189	0.098	0.116	1.570000e-01	1.570000e-01 $6.490000e-01$
ocf_at_x	532218.0	1.200000e-02 1.070000e-01	1.070000e-01	-1.358	-0.022	0.015	6.500000e-02	6.500000e-02 $5.980000e-01$
$ocf_at_chg1_x$	532218.0	-3.000000e-03	1.240000e-01	-1.078	-0.038	-0.001	3.100000e-02	3.100000e-02 1.153000e+00
mispricing_perf_x	532218.0	5.500000e-01	1.990000e-01	0.010	0.413	0.555	6.980000e-01	6.980000e-01 $9.880000e-01$
mispricing_mgmt_x	532218.0		4.850000e-01 $1.800000e-01$	0.015	0.376	0.503	6.060000e-01	6.060000e-01 $9.430000e-01$
qmj_x	532218.0	1.900000e-01	8.050000e-01	-1.703	-0.249	0.127	7.640000e-01	7.640000e-01 1.701000e+00
qmj_prof_x	532218.0	1.690000e-01	8.830000e-01	-1.704	-0.471	0.124	8.850000e-01	8.850000e-01 1.699000e+00
qmj_growth_x	532218.0	5.500000e-02	7.940000e-01	-1.701	-0.429	0.043	5.710000e-01	5.710000e-01 1.702000e+00
qmj_safety_x	532218.0		2.720000 e-01 8.670000 e-01	-1.701	-0.336	0.270	9.970000e-01	9.970000e-01 1.705000e+00

Table 5: Revised Global Factor Dataset Summary Statistics: Training Set

	count	mean	std	min	25%	20%	12%	max
$beta_60m_x$	294581.0	1.096000e+00	5.200000e-01	-1.747	8.060000e-01	1.075000e+00	1.300000e+00	.300000e+00 3.873000e+00
index	294581.0	1.362950e + 06	6.892364e + 05	1202.000	7.803110e + 05	1.500595e+06	1.905059e + 06 2.733128e + 06	2.733128e+06
mth	294581.0	1.994940e + 05	2.842530e+02	199001.000	1.992120e+05	1.995080e + 05	1.997110e+05	1.999120e+05
permno	294581.0	5.554521e+04 2.573578e+04	2.573578e+04	10010.000	3.027700e+04	6.277000e+04	7.797600e+04 9.322000e+04	9.322000e+04
permco	294581.0	1.494868e+04	8.865682e+03	4.000	7.6620000e+03	1.419100e+04	2.128600e+04 $ 5.701800e+04$	5.701800e+04
crsp_shrcd	294581.0	1.105000e+01	2.420000e-01	10.000	1.1000000 + 01	1.1000000e+01	1.1000000 + 01	1.200000e+01
crsp_exchcd	294581.0	1.860000e+00	9.660000e-01	1.000	1.0000000e+00	1.0000000e+00	3.0000000e+00	3.0000000e+00
sic	294581.0	4.670888e+03	1.847487e+03	100.000	3.330000e+03	4.813000e+03	6.153000e+03	9.997000e+03
H49	294581.0	3.098800e+01	1.313100e+01	1.000	2.1000000e+01	3.400000e+01	4.300000e+01	4.900000e+01
adjfct	294581.0	2.352000e+00	4.866000e+00	0.000	1.0000000e+00	1.0000000e+00	2.0000000e+00	2.8800000e+02
shares	294581.0	6.768500e+01	1.581650e + 02	0.542	1.463600e+01	2.761800e+01	6.018400e+01	5.160025e+03
me	294581.0	2.685416e + 03	1.012770e+04	62.992	2.917890e+02	6.017040e+02	1.685533e+03	6.024329e+05
me_company	294581.0	2.745729e+03	1.033108e+04	62.992	2.934430e+02	6.080760e + 02	1.703233e+03	6.024329e+05
prc	294581.0	4.176600e+01	7.262160e+02	0.172	1.700000e+01	2.525000e+01	3.687500e+01	7.830500e+04
prc_local	294581.0	4.176600e+01	7.262160e+02	0.172	1.700000e+01	2.525000e+01	3.687500e+01	7.830500e+04
dolvol	294581.0	1.878977e+08	7.930779e+08	0.000	1.077097e+07	3.501225e+07	1.211556e + 08	8.464558e + 10
ret	294581.0	2.400000e-02	1.420000e-01	-0.982	-4.600000e-02	1.200000e-02	7.900000e-02	9.374000e+00
ret_local	294581.0	2.400000e-02	1.420000e-01	-0.982	-4.600000e-02	1.200000e-02	7.900000e-02	9.374000e+00
ret_exc	294581.0	2.000000e-02	1.420000e-01	-0.986	-5.000000e-02	8.000000e-03	7.500000e-02	9.3700000 + 00
ret_lag_dif	294581.0	1.0000000e+00	0.0000000e+00	1.000	1.000000e+00	1.0000000e+00	1.0000000e+00	1.0000000e+00
ret_exc_lead1m	294581.0	1.000000e-02	1.340000e-01	-0.998	-5.600000e-02	5.000000e-03	6.900000e-02	6.404000e+00
niq_su_x	294581.0	-9.000000e-02	1.850000e+00	-17.871	-4.600000e-01	3.000000e-03	5.610000e-01	5.228000e+00
ret_6_1_x	294581.0	1.000000e-01	3.090000e-01	-0.838	-5.900000e-02	4.700000e-02	2.060000e-01	3.244000e+00
ret_12_1_x	294581.0	2.220000e-01	5.210000e-01	-0.907	-3.200000e-02	9.900000e-02	3.560000e-01	5.633000e+00
saleq_su_x	294581.0	2.490000e-01	1.562000e+00	-4.939	-5.680000e-01	1.380000e-01	9.630000e-01	8.854000e+00
tax_gr1a_x	294581.0	5.000000e-03	2.200000e-02	-0.132	-1.000000e-03	2.000000e-03	1.100000e-02	1.170000e-01
ni_inc8q_x	294581.0	_	3.102000e+00	0.000	0.0000000e+00	2.0000000e+00	7.0000000e+00	8.0000000e+00
$prc_highprc_252d_x$	294581.0	8.250000e-01	1.580000e-01	0.074	7.520000e-01	8.630000e-01	9.490000e-01	9.490000e-01 1.000000e+00
$ { m resff3.6.1.x} $	294581.0	-3.500000e-02	4.830000e-01	-2.692	-2.600000e-01	-2.500000e-02	2.170000e-01	2.170000e-01 1.705000e+00

	count	mean	std	mim	25%	20%	75%	max
resff3_12_1_x	294581.0	-1.500000e-02	2.450000e-01	-0.946	-1.480000e-01	-1.300000e-02	1.260000e-01	7.460000e-01
be_me_x	294581.0	5.250000e-01	4.100000e-01	0.006	2.500000e-01	4.460000e-01	6.950000e-01	1.020500e+01
debt_me_x	294581.0	5.460000e-01	1.217000e+00	0.000	3.700000e-02	2.120000e-01	5.900000e-01	5.045200e+01
at_me_x	294581.0	2.217000e+00	4.123000e+00	0.009	4.800000e-01	1.022000e+00	2.134000e+00	1.278460e + 02
$ret_60_12_x$	294581.0	8.620000e-01	1.375000e+00	-0.964	2.690000e-01	3.870000e-01	9.770000e-01	1.328900e+01
ni_me_x	294581.0	3.500000e-02	1.150000e-01	-8.106	2.000000e-02	4.700000e-02	7.200000e-02	4.830000e-01
fcf_me_x	294581.0	5.000000e-03	1.600000e-01	-5.425	-2.000000e-02	1.600000e-02	4.900000e-02	2.588000e+00
div12m_me_x	294581.0	1.400000e-02	1.900000e-02	0.000	0.0000000e+00	6.000000e-03	2.300000e-02	2.000000e-01
eqpo_me_x	294581.0	2.600000e-02	4.200000e-02	-0.001	0.0000000e+00	1.200000e-02	3.300000e-02	7.260000e-01
eqnpo_me_x	294581.0	-2.000000e-03	8.400000e-02	-1.644	-4.000000e-03	5.0000000e-03	2.700000e-02	6.710000e-01
sale_gr3_x	294581.0	8.400000e-01	2.577000e+00	-0.988	1.720000e-01	3.570000e-01	6.530000e-01	5.854700e+01
sale_gr1_x	294581.0	2.460000e-01	6.080000e-01	-0.895	3.300000e-02	1.160000e-01	2.550000e-01	1.088600e+01
ebitda_mev_x	294581.0	1.100000e-01	9.900000e-02	-1.984	6.400000e-02	1.050000e-01	1.450000e-01	1.386000e+00
sale_me_x	294581.0	1.201000e+00	1.595000e+00	0.000	3.790000e-01	7.530000e-01	1.424000e+00	5.494600e+01
ocf_me_x	294581.0	7.700000e-02	1.570000e-01	-3.829	2.400000e-02	6.800000e-02	1.190000e-01	3.793000e+00
intrinsic_value_x	294581.0	1.027543e+03	2.258605e + 03	0.141	1.283670e + 02	2.713090e+02	7.443850e+02	2.371766e+04
bev_mev_x	294581.0	5.360000e-01	3.550000e-01	0.001	2.680000e-01	5.140000e-01	7.6100000e-01	6.841000e+00
netdebt_me_x	294581.0	3.810000e-01	1.0600000e+00	-2.464	-3.000000e-02	1.210000e-01	4.720000e-01	4.390200e+01
eq_dur_x	294581.0	1.648500e+01	2.528000e+00	1.790	1.538700e+01	1.673400e+01	1.760300e+01	5.375100e+01
capex_abn_x	294581.0	6.200000e-02	6.730000e-01	-1.147	-1.480000e-01	-8.000000e-02	1.250000e-01	9.274000e+00
at_gr1_x	294581.0	3.000000e-01	7.580000e-01	-0.663	3.100000e-02	1.060000e-01	2.690000e-01	1.121900e+01
ppeinv-gr1a_x	294581.0	1.150000e-01	2.250000e-01	-0.533	3.300000e-02	5.800000e-02	1.290000e-01	2.6010000e+00
noa_at_x	294581.0	7.040000e-01	4.150000e-01	-0.509	5.610000e-01	6.720000e-01	8.0100000e-01	4.641000e+00
noa_gr1a_x	294581.0	1.380000e-01	3.540000e-01	-0.612	1.100000e-02	5.200000e-02	1.450000e-01	4.058000e+00
$lnoa_gr1a_x$	294581.0	3.000000e-02	7.000000e-02	-0.351	7.000000e-03	1.200000e-02	3.400000e-02	5.060000e-01
capx_gr1_x	294581.0	4.830000e-01	1.638000e+00	-1.337	-4.900000e-02	1.380000e-01	4.700000e-01	3.033300e+01
capx_gr2_x	294581.0	1.0100000e+00	3.574000e+00	-1.428	3.200000e-02	2.730000e-01	7.390000e-01	6.113600e+01
capx_gr3_x	294581.0	1.387000e+00 5.155000e+00	5.155000e+00	-1.199	1.450000e-01	3.660000e-01	8.780000e-01	8.780000e-01 1.076800e+02
$ chcsho_12m_x $	294581.0	6.800000e-02	2.060000e-01	-0.245	0.0000000e+00	9.000000e-03	4.300000e-02	4.300000e-02 2.658000e+00

	count	mean	std	min	25%	20%	122	max
eqnpo_12m_x	294581.0	-3.500000e-02	1.480000e-01	-1.255	-3.300000e-02	-2.000000e-03	2.700000e-02	4.210000e-01
$debt_gr3_x$	294581.0	2.550000e+00	1.542700e+01	-1.000	-3.300000e-02	1.920000e-01	7.310000e-01	3.188620e+02
inv-gr1_x	294581.0	2.290000e-01	8.790000e-01	-1.000	-4.000000e-03	9.000000e-02	2.020000e-01	1.302400e+01
inv_gr1a_x	294581.0	1.100000e-02	4.100000e-02	-0.357	-0.0000000e+00	0.00000000+000	1.500000e-02	2.630000e-01
oaccruals_at_x	294581.0	-3.400000e-02	9.600000e-02	-0.935	-6.800000e-02	-3.800000e-02	2.000000e-03	5.230000e-01
taccruals_at_x	294581.0	-5.100000e-02	1.640000e-01	-1.432	-9.400000e-02	-4.300000e-02	9.000000e-03	9.6100000e-01
cowc_gr1a_x	294581.0	1.400000e-02	6.700000e-02	-0.605	-8.000000e-03	8.0000000e-03	3.300000e-02	4.170000e-01
coa_gr1a_x	294581.0	3.900000e-02	8.300000e-02	-0.791	4.000000e-03	2.200000e-02	6.200000e-02	4.540000e-01
col-gr1a_x	294581.0	2.400000e-02	5.300000e-02	-0.471	3.000000e-03	1.500000e-02	3.800000e-02	3.440000e-01
nncoa_gr1a_x	294581.0	5.400000e-02	1.210000e-01	-1.080	7.000000e-03	2.700000e-02	8.000000e-02	6.480000e-01
$ $ ncoa_gr1a_x	294581.0	6.400000e-02	1.250000e-01	-0.943	1.200000e-02	3.300000e-02	9.400000e-02	6.630000e-01
ncol_gr1a_x	294581.0	8.0000000e-03	2.900000e-02	-0.224	-0.0000000e+00	2.000000e-03	1.100000e-02	2.190000e-01
nfna_gr1a_x	294581.0	-1.700000e-02	1.370000e-01	-0.983	-5.000000e-02	0.00000000+000	2.000000e-02	9.650000e-01
lti_gr1a_x	294581.0	9.000000e-03	4.200000e-02	-0.360	0.0000000e+00	0.00000000+000	1.000000e-03	2.940000e-01
fnl_gr1a_x	294581.0	2.400000e-02	1.260000e-01	-0.831	-1.200000e-02	0.00000000+000	4.900000e-02	9.140000e-01
$be_{gr1a.x}$	294581.0	6.900000e-02	1.440000e-01	-1.017	9.000000e-03	3.600000e-02	9.700000e-02	7.940000e-01
oaccruals_ni_x	294581.0	$294581.0 \left -1.254000e + 00 \right 4$	4.986000e+00	-60.944	-1.275000e+00	-5.450000e-01	2.200000e-02	3.908000e+01
taccruals_ni_x	294581.0	294581.0 -2.223000e + 00 8.905000e + 00	8.905000e+00	-112.262	-1.845000e+00	-6.590000e-01	1.330000e-01	3.405000e+01
netis_at_x	294581.0	1.700000e-02	2.1000000e-01	-1.171	-3.900000e-02	-1.000000e-03	3.500000e-02	1.489000e+00
eqnetis_at_x	294581.0	4.400000e-02	1.550000e-01	-0.216	-0.0000000e+00	1.000000e-03	1.300000e-02	1.389000e+00
dbnetis_at_x	294581.0	-2.400000e-02	1.460000e-01	-1.135	-4.200000e-02	-3.000000e-03	1.500000e-02	6.040000e-01
niq-be-x	294581.0	2.400000e-02	9.700000e-02	-1.913	1.500000e-02	3.000000e-02	4.900000e-02	4.290000e-01
niq-be-chg1_x	294581.0	-5.000000e-03	9.600000e-02	-1.766	-9.0000000e-03	-1.000000e-03	6.000000e-03	9.390000e-01
niq_at_x	294581.0	9.000000e-03	4.200000e-02	-0.555	3.000000e-03	1.100000e-02	2.300000e-02	1.120000e-01
niq-at_chg1_x	294581.0	294581.0 -0.000000e+00	3.500000e-02	-0.349	-4.000000e-03	-0.0000000e+00	3.000000e-03	4.710000e-01
ebit_bev_x	294581.0	1.310000e-01	8.300000e-01	-16.386	9.500000e-02	1.520000e-01	2.550000e-01	2.332000e+00
ebit_sale_x	294581.0	-1.000000e-02	1.335000e+00	-32.364	6.200000e-02	1.110000e-01	1.940000e-01	5.370000e-01
sale_bev_x	294581.0	2.200000e+00 2.665000e+00	2.665000e+00	0.000	8.240000e-01	1.530000e+00	2.507000e+00 2.745100e+01	2.745100e+01
at_turnover_x	294581.0	1.044000e+00	8.280000e-01	0.000	4.160000e-01	9.120000e-01	1.411000e+00 4.719000e+00	4.719000e+00

	count	mean	std	mim		20%	75% I	max
gp_at_x	294581.0	3.160000e-01	2.760000e-01	-0.853	1.120000e-01	2.570000e-01	4.710000e-01 1.412000e+00	00+
gp_atll_x	294581.0	3.960000e-01	3.830000e-01	-1.018	1.300000e-01	3.020000e-01	5.580000e-01 2.455000e+00	00+
ope-be-x	294581.0	2.480000e-01	3.340000e-01	-5.705	1.720000e-01	2.240000e-01	3.330000e-01 2.299000e+00	00+
ope_bell_x	294581.0	3.220000e-01	4.980000e-01	-5.420	2.020000e-01	2.470000e-01	3.970000e-01 $4.618000e+00$	00+
op_at_x	294581.0	1.470000e-01	1.280000e-01	-1.058	7.700000e-02	1.370000e-01	2.080000e-01 5.660000e-01	-01
op_atll_x	294581.0	1.830000e-01	1.990000e-01	-1.949	9.200000e-02	1.520000e-01	2.450000e-01 1.125000e+00	00+
cop_at_x	294581.0	1.890000e-01	1.500000e-01	-1.029	1.120000e-01	1.700000e-01	2.590000e-01 8.800000e-01	-01
cop_atll_x	294581.0	2.280000e-01	2.250000e-01	-1.844	1.330000e-01	1.880000e-01	3.000000e-01 1.507000e+00	00+
f_score_x	294581.0	5.173000e+00 1	1.413000e+00	1.000	5.0000000e+00	5.0000000e+00	6.0000000e+00 9.0000000e+00	00+
O_SCOTe_X	294581.0	294581.0 -2.721000e+00 2.068000e+00	2.068000e+00	-8.385	-3.800000e+00	-2.767000e+00	-2.092000e+00 1.799600e+01	+01
Z_SCOTe_X	294581.0	5.789000e+00 8.539000e+00	8.539000e+00	-7.856	2.387000e+00	3.525000e+00	5.488000e+00 9.787600e+01	+01
pi_nix_x	294581.0	1.623000e+00	6.310000e-01	0.244	1.468000e+00	1.544000e+00	1.657000e+00 9.259000e+00	00+
at_be_x	294581.0	3.946000e+00 4.895000e+00	4.895000e+00	1.023	1.596000e+00	2.137000e+00	3.387000e+00 5.215100e+01	+01
saleq_gr1_x	294581.0	2.800000e-01	7.890000e-01	-0.988	2.200000e-02	1.130000e-01	2.780000e-01 1.052800e+01	+01
opex_at_x	294581.0	8.560000e-01	7.380000e-01	0.036	3.030000e-01	7.170000e-01	1.156000e+00 $4.588000e+00$	00+
emp_gr1_x	294581.0	9.700000e-02	2.260000e-01	-1.333	2.000000e-03	6.200000e-02	1.390000e-01 1.440000e+00	00+
age_x	294581.0	2.577090e+02	3.577090e + 02 2.162820e + 02	7.000	8.1000000e+01	1.780000e+02	3.920000e+02 8.880000e+02	+02
dsale_dinv_x	294581.0	-1.700000e-02	7.120000e-01	-9.498	-4.70000e-02	3.000000e-02	1.140000e-01 3.472000e+00	00+
dsale_drec_x	294581.0	-3.200000e-02	5.320000e-01	-6.378	-1.010000e-01	2.000000e-03	9.500000e-02 3.597000e+00	00+
dgp_dsale_x	294581.0	3.200000e-02	4.350000e-01	-4.610	-4.300000e-02	2.000000e-03	6.500000e-02 $4.688000e+00$	00+
dsale_dsga_x	294581.0	2.200000e-02	2.700000e-01	-1.680	-2.400000e-02	2.000000e-03	4.000000e-02 3.562000e+00	00+
$ sale_emp_gr1_x $	294581.0	8.300000e-02	3.340000e-01	-0.841	-4.000000e-03	4.000000e-02	1.100000e-01 4.904000e+00	00+
tangibility_x	294581.0	6.350000e-01	1.730000e-01	0.111	5.600000e-01	6.440000e-01	7.300000e-01 1.390000e+00	00+
kz_index_x	294581.0	-6.962000e+00 2.127700e+01	2.127700e+01	-352.632	-5.199000e+00	-1.327000e+00	2.110000e-01 1.127800e+01	+01
cash_at_x	294581.0	1.330000e-01	1.820000e-01	0.000	1.700000e-02	5.600000e-02	1.650000e-01 9.310000e-01	-01
ni_ar1_x	294581.0	1.810000e-01	4.900000e-01	-1.898	1.000000e-02	1.240000e-01	3.310000e - 01 3.331000e + 00	00+
ni_ivol_x	294581.0	3.000000e-02	4.600000e-02	0.000	1.000000e-02	2.400000e-02	3.100000e-02 7.070000e-01	-01
earnings_variability_x 294581.0	294581.0	7.750000e-01	6.210000e-01	0.027	3.660000e-01	7.820000e-01	9.030000e-01 6.427000e+00	00+
aliq-at_x	294581.0	8.390000e-01	6.100000e-01	0.181	6.100000e-01	6.980000e-01	8.330000e-01 1.083700e+01	+01

	count	mean	std	min	25%	20%	12%	max
aliq_mat_x	294581.0	4.510000e-01	1.870000e-01	0.050	3.440000e-01	4.330000e-01	5.490000e-01	1.842000e+00
seas_1_1an_x	294581.0	2.000000e-02	1.160000e-01	-0.578	-3.700000e-02	9.000000e-03	6.900000e-02	1.407000e+00
seas_1_1na_x	294581.0	1.900000e-02	3.300000e-02	-0.160	7.000000e-03	1.400000e-02	2.800000e-02	2.910000e-01
seas_2_5an_x	294581.0	1.500000e-02	4.400000e-02	-0.276	5.000000e-03	1.200000e-02	2.600000e-02	5.270000e-01
ivol_ff3_21d_x	294581.0	2.300000e-02	1.400000e-02	0.005	1.300000e-02	1.900000e-02	2.800000e-02	1.910000e-01
ivol_capm_ $252d_x$	294581.0	2.500000e-02	1.200000e-02	0.007	1.600000e-02	2.200000e-02	3.100000e-02	1.360000e-01
ivol_capm_21d_x	294581.0	2.300000e-02	1.400000e-02	0.005	1.400000e-02	2.000000e-02	2.900000e-02	1.910000e-01
ivol_hxz4_21d_x	294581.0	2.300000e-02	1.400000e-02	0.005	1.300000e-02	1.900000e-02	2.800000e-02	1.920000e-01
rvol_21d_x	294581.0	2.500000e-02	1.500000e-02	0.006	1.500000e-02	2.200000e-02	3.100000e-02	1.990000e-01
$betabab_1260d_x$	294581.0	1.025000e+00	5.100000e-01	-0.326	7.170000e-01	9.110000e-01	1.255000e+00	3.6500000e+00
beta_dimson_21d_x	294581.0	9.630000e-01	1.780000e+00	-17.092	7.100000e-02	8.230000e-01	1.721000e+00	1.921100e+01
$turnover_126d_x$	294581.0	3.000000e-03	3.000000e-03	0.000	1.000000e-03	3.000000e-03	4.000000e-03	2.900000e-02
turnover_var_126d_x	294581.0	1.143000e+00	6.230000e-01	0.405	7.510000e-01	9.860000e-01	1.324000e+00 6.770000e+00	6.7700000+00
dolvol_126d_x	294581.0	6.536392e+06	1.464928e+07	486.102	5.446526e + 05	1.610104e+06	5.476749e+06 $1.547992e+08$	1.547992e+08
$dolvol_var_126d_x$	294581.0	1.148000e+00	6.240000e-01	0.383	7.540000e-01	1.002000e+00	1.333000e+00 6.601000e+00	6.601000e+00
ami_126d_x	294581.0	2.060000e-01	1.300000e+00	0.000	4.000000e-03	1.800000e-02	7.700000e-02	1.666870e + 02
zero_trades_21d_x	294581.0	2.430000e-01	1.301000e+00	0.000	2.0000000e-03	3.000000e-03	5.0000000e-03	5.000000e-03 2.009600e+01
$zero_trades_126d_x$	294581.0	2.490000e-01	1.223000e+00	0.000	2.0000000e-03	3.000000e-03	6.0000000e-03	6.000000e-03 1.791600e+01
$zero_trades_252d_x$	294581.0	2.580000e-01	1.216000e+00	0.000	2.0000000e-03	3.000000e-03	6.0000000e-03	1.752300e+01
rmax1_21d_x	294581.0	5.700000e-02	4.400000e-02	0.007	2.900000e-02	4.500000e-02	6.900000e-02	6.240000e-01
rskew_21d_x	294581.0	2.430000e-01	8.120000e-01	-2.818	-2.200000e-01	2.130000e-01	6.900000e-01	3.243000e+00
iskew_capm_21d_x	294581.0	2.440000e-01	7.890000e-01	-2.563	-2.120000e-01	2.130000e-01	6.800000e-01	3.091000e+00
iskew_ff3_21d_x	294581.0	1.980000e-01	7.250000e-01	-2.308	-2.310000e-01	1.730000e-01	6.050000e-01	2.8000000e+00
iskew_hxz4_21d_x	294581.0	1.780000e-01	6.920000e-01	-2.037	-2.370000e-01	1.550000e-01	5.730000e-01	2.748000e+00
coskew_21d_x	294581.0	-1.900000e-02	3.120000e-01	-1.468	-2.240000e-01	-2.200000e-02	1.840000e-01	1.154000e+00
ret_1_0x	294581.0	2.300000e-02	1.300000e-01	-0.619	-4.400000e-02	1.100000e-02	7.600000e-02	1.6000000e+00
betadown_252d_x	294581.0	9.990000e-01	7.080000e-01	-3.514	5.370000e-01	9.180000e-01		5.699000e+00
bidaskhl_21d_x	294581.0	1.000000e-02	7.000000e-03	0.003	5.000000e-03	7.000000e-03	1.200000e-02	1.790000e-01
ret_3_1_x	294581.0	4.100000e-02	1.820000e-01	-0.707	-5.200000e-02	2.000000e-02	1.160000e-01	1.160000e-01 2.077000e+00

	count	mean	std	\min	25%	20%	15%	max
$ret_9_1_x$	294581.0	294581.0 $1.590000e-01$	4.120000e-01	-0.879	-0.879 $-4.800000e-02$	7.300000e-02	2.810000e-01	2.810000e-01 4.691000e+00
ret_12_7x	294581.0	9.700000e-02	2.910000e-01	-0.797	-0.797 $-4.400000e-02$	4.200000e-02	1.960000e-01	1.960000e-01 $3.295000e+00$
$corr_1260d_x$	294581.0	3.380000e-01	1.280000e-01	-0.037	2.630000e-01	3.170000e-01	4.1100000e-01	4.110000e-01 $7.320000e-01$
$rmax5_21d_x$	294581.0	3.300000e-02	2.200000e-02	0.005	1.900000e-02	2.800000e-02	4.100000e-02	4.100000e-02 $2.990000e-01$
rmax5_rvol_21d_x	294581.0	294581.0 1.265000e+00	4.620000e-01	0.234	9.530000e-01	1.190000e+00	1.522000e+00 3.545000e+00	3.545000e+00
ni_be_x	294581.0	6.800000e-02	3.810000e-01	-7.304	6.100000e-02	1.110000e-01	1.650000e-01	1.650000e-01 8.470000e-01
ocf_at_x	294581.0	6.500000e-02	1.280000e-01	-1.240	1.900000e-02	7.100000e-02	1.260000e-01	1.260000e-01 $4.950000e-01$
$ocf_at_chg1_x$	294581.0	6.000000e-03	1.090000e-01	-0.852	-2.600000e-02	-2.600000e-02 -0.0000000e+00	3.000000e-02	3.000000e-02 1.036000e+00
mispricing_perf_x	294581.0	5.780000e-01	1.820000e-01	0.021	4.540000e-01	5.790000e-01	7.130000e-01	7.130000e-01 9.610000e-01
mispricing_mgmt_x	294581.0	4.900000e-01	1.790000e-01	0.052	3.680000e-01	4.950000e-01	6.230000e-01	6.230000e-01 $8.840000e-01$
qmj_x	294581.0	2.170000e-01	7.590000e-01	-1.698	-1.698 -7.800000e-02	1.140000e-01	7.470000e-01	7.470000e-01 $1.698000e+00$
qmj_prof_x	294581.0	3.300000e-01	9.150000e-01	-1.698	-1.698 -3.660000e-01	4.030000e-01	1.126000e+00	1.126000e+00 1.698000e+00
qmj-growth-x	294581.0	7.200000e-02	7.390000e-01	-1.698	-1.698 -2.810000e-01	3.500000e-02	5.130000e-01	5.130000e-01 1.698000e + 00
qmj_safety_x	294581.0	2.230000e-01	8.930000e-01	-1.698	-1.698 $-4.430000e-01$	2.640000e-01	9.700000e-01	9.700000e-01 1.701000e+00

Table 6: Revised Global Factor Dataset Summary Statistics: Validation Set

	count	mean	std	min	25%	20%	75%	max
$beta_60m_x$	531461.0	1.143000e+00	6.650000e-01	-0.762	7.160000e-01	1.081000e+00	1.428000e+00 4.912000e+00	4.912000e+00
index	531461.0	531461.0 1.639772e+06	8.608909e+05	1495.000	8.349320e+05	1.872142e+06	2.405852e+06 $2.739927e+06$	2.739927e+06
mth	531461.0	2.009810e + 05	6.308130e + 02	200001.000	2.004070e+05	2.010030e+05	2.015100e+05 2.020120e+05	2.020120e+05
permno	531461.0	531461.0 6.249862e+04	2.927243e+04	10012.000	3.294200e+04	7.752600e+04	8.648900e+04 9.343600e+04	9.343600e+04
permco	531461.0	2.536554e+04	1.682836e+04	7.000	1.205800e+04	2.098100e+04	4.099800e+04 5.766700e+04	5.766700e+04
crsp_shrcd	531461.0	1.110400e+01	3.060000e-01	10.000	1.1000000e+01	1.1000000e+01	1.1000000e+01	1.200000e+01
crsp_exchcd	531461.0	1.820000e+00	9.770000e-01	1.000	1.0000000e+00	1.0000000e+00	3.0000000e+00	3.0000000e+00
sic	531461.0	4.731309e+03	1.973094e+03	100.000	3.312000e+03	4.813000e+03	6.2110000e+03	9.999000e+03
H49	531461.0	3.088900e+01	1.275800e+01	1.000	2.1000000e+01	3.4000000e+01	4.200000e+01	4.900000e+01
adjfct	531461.0	1.190000e+00	1.201000e+00	0.000	1.0000000e+00	1.0000000e+00	1.000000e+00	1.120000e+02
shares	531461.0	2.183130e+02	6.077150e+02	1.009	3.797500e+01	7.258200e+01	1.709880e+02	2.920640e+04
me	531461.0	9.114197e+03	3.153046e+04	178.093	9.360740e+02	2.009547e+03	5.793820e+03	2.255969e+06
me_company	531461.0	9.284112e+03	3.274782e+04	178.093	9.401740e+02	2.023393e+03	5.869141e+03	2.255969e+06
prc	531461.0	6.448400e+01	1.368034e+03	0.120	1.815000e+01	3.059000e+01	4.950000e+01	1.4160000e+05
prc_local	531461.0	531461.0 6.448400e+01	1.368034e+03	0.120	1.815000e+01	3.059000e+01	4.950000e+01	1.4160000e+05
dolvol	531461.0	1.430663e + 09	5.379275e+09	0.000	1.191406e+08	3.596018e+08	1.153573e+09	8.441730e+11
ret	531461.0	1.800000e-02	1.510000e-01	-0.984	-4.800000e-02	1.200000e-02	7.300000e-02	1.988400e+01
ret_local	531461.0	1.800000e-02	1.510000e-01	-0.984	-4.800000e-02	1.200000e-02	7.300000e-02	1.988400e+01
ret_exc	531461.0	1.600000e-02	1.510000e-01	-0.985	-4.900000e-02	1.000000e-02	7.200000e-02	1.988200e+01
ret_lag_dif	531461.0	1.0000000e+00	0.0000000e+00	1.000	1.0000000e+00	1.0000000e+00	1.0000000e+00	1.0000000e+00
ret_exc_lead1m	531461.0	7.0000000e-03	1.370000e-01	-1.001	-5.400000e-02	7.0000000e-03	6.700000e-02	4.1000000e+000
niq_su_x	531461.0	-9.700000e-02	1.914000e+00	-50.846	-6.530000e-01	-1.100000e-02	6.410000e-01	2.019500e+01
ret_6_1_x	531461.0	8.200000e-02	3.500000e-01	-0.917	-7.900000e-02	4.600000e-02	1.920000e-01	8.412000e+00
ret_12_1x	531461.0	1.800000e-01	5.820000e-01	-0.973	-8.800000e-02	8.600000e-02	3.200000e-01	1.223100e+01
saleq_su_x	531461.0	1.380000e-01	1.682000e+00	-16.096	-7.990000e-01	1.300000e-01	1.0620000e+00	8.648000e+00
tax_gr1a_x	531461.0	2.0000000e-03	2.800000e-02	-0.216	-4.000000e-03	1.000000e-03	9.000000e-03	2.050000e-01
ni_inc8q_x	531461.0		3.092000e+00	0.000	0.00000000+00	2.0000000e+00	6.0000000e+00	8.0000000e+00
$prc_highprc_252d_x$	531461.0		1.820000e-01	0.017	7.290000e-01	8.640000e-01	9.510000e-01 1.000000e+00	1.0000000e+000
$ { m resff3.6.1.x} $	531461.0	-2.900000e-02	4.950000e-01	-2.363	-2.870000e-01	-2.200000e-02	2.540000e-01 1.833000e+00	1.833000e+00

	count	mean	$ \operatorname{std} $	min	25%	20%	75% max
resff3_12_1_x	531461.0	-1.30000e-02	2.530000e-01	-1.046	-1.600000e-01	-1.200000e-02	1.440000e-01 $7.740000e-01$
be_me_x	531461.0	5.420000e-01	5.170000e-01	0.005	2.440000e-01	4.390000e-01	6.890000e-01 $2.516300e+01$
debt_me_x	531461.0	5.270000e-01	1.260000e+00	0.000	4.300000e-02	2.080000e-01	5.440000e-01 $ 6.550600e+01 $
at_me_x	531461.0	1.997000e+00 3.	3.652000e+00	0.010	4.730000e-01	9.520000e-01	$2.011000e + 00 \left 1.923120e + 02 \right $
$ret_60_12_x$	531461.0	7.400000e-01	1.402000e+00	-0.996	1.350000e-01	3.530000e-01	9.310000e-01 1.791400e+01
ni_me_x	531461.0	1.900000e-02	2.300000e-01	-18.929	1.300000e-02	4.200000e-02	6.600000e-02 $9.920000e-01$
fcf_me_x	531461.0	2.900000e-02	1.710000e-01	-7.856	2.000000e-03	3.700000e-02	7.200000e-02 $4.202000e+00$
div12m_me_x	531461.0	1.200000e-02	1.900000e-02	0.000	0.0000000e+00	2.000000e-03	1.900000e-02 $4.020000e-01$
eqpo_me_x	531461.0	3.400000e-02	5.400000e-02	-0.000	1.000000e-03	1.600000e-02	4.500000e-02 1.725000e+00
eqnpo_me_x	531461.0	9.000000e-03	9.600000e-02	-3.911	-3.000000e-03	9.0000000e-03	3.800000e-02 1.443000e+00
sale_gr3_x	531461.0	7.190000e-01	2.545000e+00	-1.000	6.800000e-02	2.950000e-01	6.090000e-01 8.620400e+01
$sale_gr1_x$	531461.0	1.910000e-01	6.470000e-01	-0.996	4.000000e-03	8.700000e-02	2.090000e-01 1.370600e+01
ebitda_mev_x	531461.0	9.300000e-02	1.190000e-01	-5.587	5.500000e-02	9.000000e-02	1.290000e-01 1.983000e+00
sale_me_x	531461.0	9.930000e-01 1.	1.6060000e+00	0.000	2.740000e-01	5.560000e-01	$1.105000e{+00} \left 7.507500e{+01} \right $
ocf_me_x	531461.0	9.300000e-02	1.560000e-01	-4.343	3.900000e-02	7.500000e-02	1.280000e-01 5.711000e+00
intrinsic_value_x	531461.0	3.673158e + 039	9.322883e+03	0.515	2.746770e+02	6.616840e + 02	2.401336e+03 1.130984e+05
bev_mev_x	531461.0	5.450000e-01	4.310000e-01	0.001	2.580000e-01	4.990000e-01	7.490000e-01 1.692600e+01
netdebt_me_x	531461.0	3.480000e-01	1.073000e+00	-3.496	-5.200000e-02	1.000000e-01	4.220000e-01 $ 5.866300e+01$
eq_dur_x	531461.0	1.683500e+01 5	5.240000e+00	1.053	1.565800e+01	1.686700e + 01	1.783200e+01 3.430350e+02
capex_abn_x	531461.0	4.500000e-02	6.920000e-01	-1.028	-2.420000e-01	-6.700000e-02	1.680000e-01 $1.196300e+01$
at_gr1_x	531461.0	2.660000e-01	1.181000e+00	-0.740	3.000000e-03	7.700000e-02	2.080000e-01 $3.163800e+01$
ppeinv-gr1a_x	531461.0	7.400000e-02	1.740000e-01	-0.449	1.000000e-02	4.000000e-02	8.300000e-02 3.079000e+00
noa_at_x	531461.0	6.120000e-01	4.870000e-01	-1.152	4.460000e-01	6.240000e-01	7.550000e-01 1.038800e+01
noa_gr1a_x	531461.0	1.040000e-01	4.220000e-01	-0.737	-1.300000e-02	3.300000e-02	1.020000e-01 1.075200e+01
lnoa_gr1a_x	531461.0	3.500000e-02	1.020000e-01	-0.578	-4.000000e-03	1.200000e-02	4.500000e-02 $7.540000e-01$
capx_gr1_x	531461.0	3.700000e-01	1.543000e+00	-1.021	-1.390000e-01	1.030000e-01	4.080000e-01 $3.277600e+01$
capx_gr2_x	531461.0	7.800000e-01 3	3.2100000e+00	-1.010	-1.390000e-01	2.160000e-01	6.650000e-01 $7.697200e+01$
capx_gr3_x	531461.0	1.128000e+00	4.517000e+00	-1.000	-1.030000e-01	2.930000e-01	8.460000e-01 1.128460e+02
chcsho_12m_x	531461.0	4.600000e-02	2.130000e-01	-0.227	-9.000000e-03	7.000000e-03	2.600000e-02 8.477000e+00

	count	mean	std	min	25%	20%	75%	max
eqnpo_12m_x	531461.0	-2.000000e-02	1.340000e-01	-2.249	-2.100000e-02	1.000000e-03	3.300000e-02	3.470000e-01
$debt_gr3_x$	531461.0	531461.0 4.136000e+00 2.614000e+01	2.614000e+01	-1.000	-9.000000e-02	1.730000e-01	7.340000e-01	4.310000e+02
inv-gr1_x	531461.0	1.820000e-01	8.460000e-01	-1.000	-1.300000e-02	7.700000e-02	1.570000e-01	1.698100e+01
inv-gr1a_x	531461.0	6.000000e-03	2.900000e-02	-0.288	-0.0000000e+00	0.0000000e+00	9.0000000e-03	2.080000e-01
oaccruals_at_x	531461.0	-5.600000e-02	1.000000e-01	-2.264	-8.100000e-02	-4.400000e-02	-1.400000e-02	3.810000e-01
taccruals_at_x	531461.0	-6.900000e-02	1.810000e-01	-2.480	-1.150000e-01	-4.900000e-02	-4.000000e-03	1.294000e+00
cowc_gr1a_x	531461.0	3.000000e-03	5.500000e-02	-0.605	-1.400000e-02	3.000000e-03	2.0000000e-02	3.410000e-01
coa_gr1a_x	531461.0	2.000000e-02	6.500000e-02	-0.777	-2.000000e-03	1.400000e-02	3.900000e-02	4.190000e-01
col_gr1a_x	531461.0	1.700000e-02	5.000000e-02	-0.485	-2.000000e-03	1.100000e-02	3.0000000e-02	3.830000e-01
nncoa_gr1a_x	531461.0	4.100000e-02	1.340000e-01	-1.884	-8.000000e-03	2.000000e-02	7.000000e-02	7.690000e-01
$ncoa_gr1a_x$	531461.0	4.900000e-02	1.420000e-01	-1.884	-4.000000e-03	2.600000e-02	8.100000e-02	7.490000e-01
ncol_gr1a_x	531461.0	7.000000e-03	3.600000e-02	-0.360	-2.000000e-03	2.000000e-03	1.500000e-02	3.340000e-01
nfna_gr1a_x	531461.0	-1.400000e-02	1.460000e-01	-1.108	-4.800000e-02	-0.0000000e+00	2.700000e-02	1.384000e+00
lti_gr1a_x	531461.0	5.000000e-03	3.500000e-02	-0.231	0.00000000+000	0.0000000e+00	1.000000e-03	2.570000e-01
fnl-gr1a_x	531461.0	2.200000e-02	1.270000e-01	-1.230	-1.400000e-02	0.0000000e+00	4.300000e-02	1.130000e+00
be_gr1a_x	531461.0	4.500000e-02	1.460000e-01	-2.072	1.000000e-03	2.800000e-02	8.100000e-02	8.560000e-01
oaccruals_ni_x	531461.0	531461.0 -2.033000e + 00 5	5.721000e+00	-71.442	-1.715000e+00	-7.240000e-01	-2.420000e-01	1.429400e+01
taccruals_ni_x	531461.0	531461.0 -2.755000e+00 9	9.625000e+00	-131.510	-2.469000e+00	-8.370000e-01	-6.300000e-02	6.728600e+01
netis_at_x	531461.0	-2.800000e-02	2.270000e-01	-1.368	-8.500000e-02	-1.800000e-02	1.300000e-02	1.593000e+00
eqnetis_at_x	531461.0	2.100000e-02	1.510000e-01	-0.351	-1.500000e-02	0.00000000+00	7.000000e-03	1.489000e+00
dbnetis_at_x	531461.0	-4.800000e-02	1.610000e-01	-1.362	-6.200000e-02	-6.000000e-03	0.0000000e+00	5.670000e-01
niq-be_x	531461.0	1.800000e-02	1.130000e-01	-2.022	9.000000e-03	2.600000e-02	4.600000e-02	6.990000e-01
niq-be-chg1_x	531461.0	-2.000000e-03	1.130000e-01	-2.004	-1.100000e-02	-1.000000e-03	1.000000e-02	1.228000e+00
niq-at_x	531461.0	6.0000000e-03	4.700000e-02	-0.667	2.000000e-03	1.000000e-02	2.200000e-02	1.820000e-01
niq-at_chg1_x	531461.0	1.000000e-03	4.500000e-02	-0.455	-5.000000e-03	0.00000000+00	5.000000e-03	8.410000e-01
ebit_bev_x	531461.0	5.600000e-02 1	1.346000e+00	-41.056	7.400000e-02	1.310000e-01	2.330000e-01	2.800000e+00
ebit_sale_x	531461.0	-2.370000e-01		-185.045	5.700000e-02	1.170000e-01	2.090000e-01	6.150000e-01
sale_bev_x	531461.0	2.041000e+00		0.000	6.210000e-01	1.261000e+00	2.224000e+00	3.887100e+01
at_turnover_x	531461.0	8.410000e-01	7.220000e-01	0.000	3.210000e-01	6.760000e-01	1.141000e+00	1.141000e+00 4.267000e+00

	count	mean	std	mim	25%	20%	75% max
gp_at_x	531461.0	2.740000e-01	2.480000e-01	-1.169	1.0000000e-01	2.390000e-01	3.990000e-01 1.293000e+00
gp_atll_x	531461.0	3.240000e-01	3.250000e-01	-1.904	1.130000e-01	2.680000e-01	4.630000e-01 2.788000e+00
ope-be-x	531461.0	2.340000e-01	4.400000e-01	-8.815	1.450000e-01	2.040000e-01	3.170000e-01 3.725000e+00
ope_bell_x	531461.0	2.740000e-01	5.480000e-01	-13.628	1.680000e-01	2.280000e-01	3.610000e-01 4.420000e+00
op_at_x	531461.0	1.320000e-01	1.220000e-01	-1.224	6.700000e-02	1.260000e-01	1.940000e-01 $5.380000e-01$
op_atll_x	531461.0	1.470000e-01	2.450000e-01	-6.946	7.400000e-02	1.400000e-01	2.240000e-01 1.018000e+00
cop_at_x	531461.0	1.910000e-01	1.510000e-01	-0.924	1.070000e-01	1.760000e-01	2.650000e-01 1.940000e+00
cop_atll_x	531461.0	2.190000e-01	2.240000e-01	-3.834	1.190000e-01	1.960000e-01	3.020000e-01 1.923000e+00
f_score_x	531461.0	5.234000e+00	1.422000e+00	1.000	4.000000e+00	5.000000e+00	6.000000e+00 $9.000000e+00$
O_SCOFe_X	531461.0	531461.0 -2.944000e+00 2	2.326000e+00	-9.387	4.243000e+00	-3.020000e+00	-2.185000e+00 2.287000e+01
Z_SCOTe_X	531461.0	5.413000e+00 9	9.607000e+00	-37.336	1.997000e+00	3.275000e+00	5.180000e+00 1.744240e+02
pi_nix_x	531461.0	1.519000e+00	6.770000e-01	0.106	1.334000e+00	1.488000e+00	1.589000e+00 1.989400e+01
at_be_x	531461.0	3.707000e+00 $4.571000e+00$	4.571000e+00	1.023	1.582000e+00	2.157000e+00	3.430000e+00 5.963100e+01
$saleq_gr1_x$	531461.0	2.020000e-01	7.530000e-01	-1.000	-7.000000e-03	8.500000e-02	2.150000e-01 1.574800e+01
opex_at_x	531461.0	7.050000e-01	6.560000e-01	0.003	2.430000e-01	5.380000e-01	9.460000e-01 4.437000e+00
emp_gr1_x	531461.0	7.300000e-02	2.000000e-01	-1.165	-1.000000e-02	5.100000e-02	1.190000e-01 1.442000e+00
age_x	531461.0	3.164250e + 02 2.392710e + 02	2.392710e+02	10.000	1.300000e+02	2.440000e+02	4.400000e+02 1.115000e+03
dsale_dinv_x	531461.0	-2.300000e-02	7.480000e-01	-19.478	-6.100000e-02	2.500000e-02	9.800000e-02 5.598000e+00
dsale_drec_x	531461.0	-2.200000e-02	5.420000e-01	-7.400	-1.120000e-01	4.000000e-03	1.160000e-01 7.638000e+00
dgp_dsale_x	531461.0	3.700000e-02	4.720000e-01	-5.970	-4.200000e-02	5.000000e-03	6.500000e-02 7.188000e+00
dsale_dsga_x	531461.0	2.300000e-02	2.880000e-01	-2.001	-4.200000e-02	3.000000e-03	5.400000e-02 $6.964000e+00$
$sale_emp_gr1_x$	531461.0	7.800000e-02	3.670000e-01	-0.956	-3.000000e-02	3.900000e-02	1.070000e-01 7.027000e+00
$tangibility_x$	531461.0	5.840000e-01	2.060000e-01	0.002	4.580000e-01	6.190000e-01	7.020000e-01 1.685000e+00
kz_index_x	531461.0	$531461.0 \left -1.475800e + 01 \right 5.964000e + 01$	5.964000e+01	-1723.572	9.717000e+00	-2.274000e+00	2.400000e-01 8.730000e+01
cash_at_x	531461.0	1.680000e-01	2.030000e-01	0.000	2.900000e-02	8.300000e-02	2.240000e-01 $9.800000e-01$
ni_ar1_x	531461.0	1.460000e-01	5.670000e-01	-3.964	-1.330000e-01	1.060000e-01	3.650000e-01 9.144000e+00
ni_ivol_x	531461.0	4.200000e-02	8.300000e-02	0.000	1.000000e-02	2.300000e-02	3.700000e-02 1.757000e+00
earnings_variability_x 531461.0	531461.0	1.119000e+00	1.132000e+00	0.044	5.490000e-01	8.810000e-01	1.192000e+00 1.145300e+01
aliq_at_x	531461.0	7.410000e-01	9.790000e-01	0.104	5.050000e-01	6.250000e-01	7.440000e-01 2.804000e+01

	count	mean	std	min	25%	20%	75%	max
aliq_mat_x	531461.0	3.770000e-01	1.890000e-01	0.027	2.380000e-01	3.610000e-01	4.780000e-01	2.246000e+00
seas_1_1an_x	531461.0	1.600000e-02	1.280000e-01	-0.670	-4.500000e-02	7.000000e-03	6.800000e-02	1.824000e+00
seas_1_1na_x	531461.0	1.600000e-02	3.900000e-02	-0.235	-2.000000e-03	1.200000e-02	3.0000000e-02	3.870000e-01
seas_2_5an_x	531461.0	1.500000e-02	5.600000e-02	-0.297	-9.000000e-03	1.100000e-02	3.500000e-02	6.340000e-01
ivol_ff3_21d_x	531461.0	2.000000e-02	1.500000e-02	0.002	1.000000e-02	1.600000e-02	2.400000e-02	2.280000e-01
$ivol_capm_252d_x$	531461.0	2.300000e-02	1.300000e-02	0.005	1.400000e-02	2.000000e-02	2.800000e-02	1.410000e-01
ivol_capm_21d_x	531461.0	2.100000e-02	1.600000e-02	0.002	1.100000e-02	1.600000e-02	2.6000000e-02	2.420000e-01
ivol_hxz4_21d_x	531461.0	2.000000e-02	1.500000e-02	0.002	1.000000e-02	1.500000e-02	2.400000e-02	2.310000e-01
rvol_21d_x	531461.0	2.500000e-02	1.800000e-02	0.002	1.400000e-02	2.000000e-02	3.100000e-02	2.520000e-01
$betabab_1260d_x$	531461.0	1.071000e+00	4.730000e-01	-0.029	7.670000e-01	1.029000e+00	1.303000e+00	3.831000e+00
beta_dimson_21d_x	531461.0	1.175000e+00	1.449000e+00	-13.953	4.250000e-01	1.041000e+00	1.786000e+00	1.954200e+01
turnover_126d_x	531461.0	9.0000000e-03	8.000000e-03	0.000	4.000000e-03	7.000000e-03	1.200000e-02	2.800000e-01
turnover_var_126d_x	531461.0	7.410000e-01	4.240000e-01	0.280	4.880000e-01	6.180000e-01	8.450000e-01	7.5100000e+00
dolvol_126d_x	531461.0	5.434158e+07	9.867098e+07	2317.257	5.641453e+06	1.690584e + 07	5.411359e+07	1.038495e+09
dolvol_var_126d_x	531461.0	7.390000e-01	4.300000e-01	0.262	4.840000e-01	6.190000e-01	8.520000e-01	8.520000e-01 8.104000e+00
ami_126d_x	531461.0	2.500000e-02	6.470000e-01	0.000	0.0000000e+00	1.000000e-03	4.000000e-03	4.000000e-03 9.796700e+01
$zero_trades_21d_x$	531461.0	1.100000e-02	2.260000e-01	0.000	1.000000e-03	2.000000e-03	3.0000000e-03	3.000000e-03 1.575900e+01
zero_trades_126d_x	531461.0	1.300000e-02	2.210000e-01	0.000	1.000000e-03	2.000000e-03	3.0000000e-03	1.400900e+01
$zero_trades_252d_x$	531461.0	1.500000e-02	2.330000e-01	0.000	1.000000e-03	2.000000e-03	3.000000e-03	1.359200e+01
rmax1_21d_x	531461.0	5.6000000e-02	5.000000e-02	0.003	2.700000e-02	4.100000e-02	6.6000000e-02	9.000000e-01
rskew_21d_x	531461.0	1.350000e-01	9.260000e-01	-3.581	-3.520000e-01	1.140000e-01	5.940000e-01	3.808000e+00
iskew_capm_21d_x	531461.0	1.560000e-01	9.940000e-01	-3.566	-3.640000e-01	1.340000e-01	6.520000e-01	3.715000e+00
iskew_ff3_21d_x	531461.0	1.300000e-01	8.940000e-01	-3.120	-3.550000e-01	1.110000e-01	5.920000e-01	3.456000e+00
iskew_hxz4_21d_x	531461.0	1.200000e-01	8.400000e-01	-3.081	-3.460000e-01	1.020000e-01	5.650000e-01	3.276000e+00
coskew_21d_x	531461.0	-3.000000e-03	3.090000e-01	-1.128	-2.090000e-01	-7.000000e-03	2.000000e-01	1.347000e+00
ret_1_0x	531461.0	1.700000e-02	1.360000e-01	-0.724	-4.70000e-02	1.100000e-02	7.200000e-02	2.176000e+00
betadown_252d_x	531461.0	1.087000e+00	5.640000e-01	-1.626	7.370000e-01	1.024000e+00	1.371000e+00	3.822000e+00
bidaskhl_21d_x	531461.0	9.000000e-03	6.0000000-03	0.001	5.000000e-03	7.000000e-03	1.100000e-02	1.540000e-01
ret_3_x	531461.0	3.300000e-02	1.990000e-01	-0.831	-6.100000e-02	2.000000e-02	1.090000e-01	1.090000e-01 3.342000e+00

	count	mean	std	$ \min $	25%	20%	12%	max
ret_9_1x	531461.0	531461.0 $1.300000e-01$	4.620000e-01	-0.955	-0.955 $ -8.700000e-02 $	6.700000e-02		2.590000e-01 9.274000e+00
ret_12_7x	531461.0	7.800000e-02	3.330000e-01	-0.906	-0.906 $-7.500000e-02$	3.900000e-02	1.840000e-01	1.840000e-01 $8.509000e+00$
corr_1260d_x	531461.0	4.620000e-01	1.510000e-01	-0.010	3.610000e-01	4.500000e-01	5.780000e-01	5.780000e-01 8.220000e-01
rmax5_21d_x	531461.0	3.200000e-02	2.400000e-02	0.002	1.700000e-02	2.500000e-02	3.900000e-02	3.900000e-02 $3.380000e-01$
rmax5_rvol_21d_x	531461.0	531461.0 1.195000e+00	4.900000e-01	0.112	8.620000e-01	1.117000e+00	1.435000e+00 4.329000e+00	4.329000e+00
ni_be_x	531461.0	5.000000e-02	4.660000e-01	-10.754	3.800000e-02	9.600000e-02	1.630000e-01	1.630000e-01 1.451000e+00
ocf_at_x	531461.0	7.400000e-02	1.290000e-01	-1.818	3.100000e-02	8.100000e-02	1.330000e-01	1.330000e-01 $3.960000e-01$
$ ocf_at_chg1_x $	531461.0	7.000000e-03	1.060000e-01	-0.912	-0.912 $-2.300000e-02$	1.000000e-03	2.700000e-02	2.700000e-02 $1.390000e+00$
mispricing_perf_x	531461.0	5.800000e-01	1.740000e-01	0.031	4.650000e-01	5.890000e-01	7.090000e-01	7.090000e-01 $9.240000e-01$
mispricing_mgmt_x	531461.0	4.940000e-01	1.650000e-01	0.065	3.850000e-01	5.040000e-01	6.150000e-01	6.150000e-01 8.790000e-01
qmj_x	531461.0	2.090000e-01	8.600000e-01	-1.698	-1.698 -3.550000e-01	1.780000e-01	9.070000e-01	9.070000e-01 $1.698000e+00$
qmj_prof_x	531461.0	3.650000e-01	9.230000e-01	-1.698	-1.698 -3.270000e-01	4.870000e-01	1.161000e+00 1.698000e+00	1.698000e+00
$ qmj_growth_x $	531461.0	7.800000e-02	8.590000e-01	-1.698	-1.698 -5.550000e-01	1.120000e-01	7.300000e-01	7.300000e-01 $1.698000e+00$
$ qmj_safety_x $	531461.0	1.040000e-01	9.040000e-01	-1.698	-1.698 -6.120000e-01	1.430000e-01	8.510000e-01	8.510000e-01 1.708000e+00

Table 7: Revised Global Factor Dataset Summary Statistics: Testing Set