

QUANTITATIVE PORTFOLIO MANAGEMENT CONFERENCE

Linear Factor Models: Structure, Estimation, and Factor Selection

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Agenda

- Linear factor models
 - Motivation
 - Estimation
 - Factors and loadings
 - Equity vs. fixed income models
- Factor / model selection
 - Quantitative measures
 - In-sample analysis
 - Out-of-sample analysis
 - Qualitative measures





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Linear Factor Models



Linear Factor Models – Motivation

- Portfolio managers want to
 - Estimate the total risk of their portfolios
 - Analyze portfolio risk in various dimensions: Sources of risk, contributions
- A simple idea: Estimate the risk of individual securities and then combine them
 - Individual securities are too noisy
 - Portfolios are generally composed of a large number of securities
- Factor model: Separate the return into systematic and idiosyncratic components
 - Idiosyncratic component the noise diversifies away in portfolios
 - Reduce dimensionality, a more robust structure to estimate future properties of the portfolio return distribution
- Two components in the estimation of portfolio risk (volatility vs. tail risk)
 - Construction of the factor model $r = L * F + \varepsilon$
 - Estimation of the factor correlations and volatilities (covariance matrix) $\sigma_r^2 = L^T * \sum L + \Omega$

Linear Factor Models – Introduction

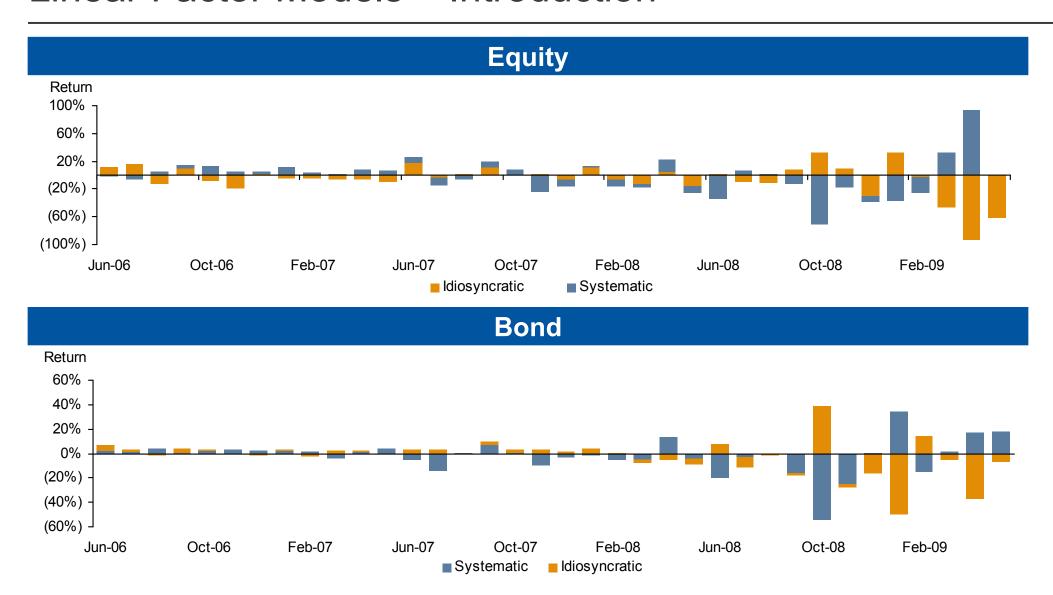
- Decompose security returns into a systematic and an idiosyncratic component
 - Systematic return: Generated by movements in common risk factors
 - Factors (F): Drivers of return
 - Loadings (L): Sensitivities to risk factors
 - Idiosyncratic return: Security-specific component unexplained by systematic factors
- An example: GM equity and bond systematic and idiosyncratic returns from June 2006 to June 2009
 - Systematic component is a bigger part of total return for the bond
 - Equity return is more volatile
 - Similar behavior in both issues moving toward bankruptcy

Systematic Ret / Idiosyncratic Ret					
	Full History	Before Sep-08	Monthly Volatility		
Equity	0.96	1.00	18.4%		
Bond	1.24	1.74	9.1%		

Note: The source for all figures in this presentation is Barclays Capital.



Linear Factor Models – Introduction





Linear Factor Models – Estimation

Choice of factors determines the characteristics of loadings and vice versa

$$r = L * F + \varepsilon$$

Model	Known	Estimated
Cross-sectional	r,L	F
Time series	r,F	L
Statistical	r	L,F

Heteroskedasticity in residual volatilities: Weighted regressions

	Cross Sectional	Time Series
Input set	Security-specific loadings – rich	Factor history – easier to access
# of parameters	(no. of factors) * (no. of months)	(no. of securities) * (no. of factors)
Interpretation	Cleaner interpretation of loadings and univariate factors	Straightforward interpretation of factors
Types of factors	Fundamental/Technical – Estimated, Classification – Estimated	Macroeconomic, Classification – Observed Fundamental/Technical – Constructed
Advantages	More dynamic in nature: loadings can adjust very fast	Direct implication with respect to time series characteristics of factors and residuals



Linear Factor Models – Factors

- Factors can be
 - Observed: Change in interest rates, equity market return S&P500
 - Estimated: Through cross-sectional regressions $\hat{F} = (L'L)^{-1}L'r$
 - Constructed: Factor mimicking portfolios (Fama and French) $\sum_{\substack{i \in Small \\ Cap}} w_i r_i \sum_{\substack{i \in L \text{ arg } e \\ Cap}} w_i r_i$
- Underlying variables
 - Macroeconomic: Macro variables that represent the state of the economy
 - Change in interest rates, industrial production
 - Fundamental and technical: Investment themes for portfolio managers
 - Size, value, momentum
 - Classification: Widely used in various models
 - Industry, country, rating
 - Statistical: No direct economic interpretation; may be remapped to known factors
 - First three PCs for rates: Level, slope, curvature. First PC in equities: Market



Linear Factor Models – Loadings

- Loadings can be
 - Analytical: Coming out of a pricing model (bond durations, option greeks)
 - Used primarily in "structured" assets, e.g., IG credit bonds
 - The market translates the risk-free rate and the quoted spread into the bond price using a pricing model; analytics can be generated

$$OAD = -100 * \frac{1}{P} * \frac{\partial P}{\partial r} \approx \frac{P_{-25bps} - P_{+25bps}}{P_0^{Full} * 2 * .25}$$

- Empirical: Estimated using statistical techniques (equity market beta)
 - Used when market dynamics (which are hard to model) are more important than fundamental relationships
 - Asset and factor returns must be sufficiently "invariant"

$$\hat{L} = F^T r / F^T F$$

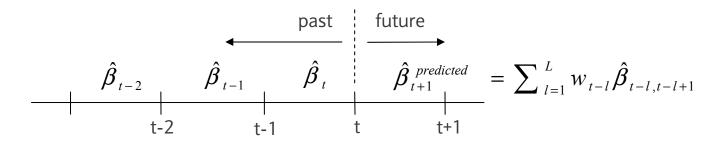
- Observed: Security characteristics used in cross-sectional models (E/P, size)
- Dummy: Used with classification factors (industry, country dummies)



Linear Factor Models – Loadings

- Standardized loadings
 - Implies a relative view, de-trends the loading (e.g., size factor)
- $L^{STD} = \frac{L \mu}{\sigma}$

- Helps produce a more invariant factor
- Use of mixed frequency data to estimate the empirical loadings
 - High frequency data to produce dynamic sensitivities: More robust, yet responsive



- Linear factor models do not restrict non-linear relationships
 - Non-linear transformations for factor loadings

$$r = g(L) * F + \varepsilon$$





Linear Factor Models – Fixed Income vs. Equity

- Fixed income
 - Generally structured assets; pricing model determines the factor structure
 - Variables in the pricing model: Interest rates, credit spread, implied volatility
 - Analytical loadings are widely used: e.g., Durations

Credit Bond
$$r_{BOND} = r_{carry} + \sum_{i=1}^{N} KRD_i*(-\Delta KR_i) + Vega*F^{\sigma} + OASD*F^{OASD} + \varepsilon$$

Carry Interest rate Volatility Credit Spread

- Equity
 - Pricing models are not widely used in practice: High uncertainty about dividends, companies with no dividends
 - Empirical sensitivities are more common
 - Relatively smaller explanatory power

Common Stock
$$r_{EQ} = \beta * F^{IND} + \sum_{i=1}^{N} \gamma_i * F_i^{FUND} + \varepsilon$$
Industry Fund & Tech





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Factor / Model Selection



Factor / Model Selection – Motivation

- Goal: From a large set of n potential factors, select a subset of k factors in an optimal way to
 - Produce accurate estimates of portfolio risk
 - Across different market conditions and for a diverse set of portfolios
 - Facilitate intuitive analysis of total risk: Decomposition, exposures
- Choice of a factor cannot be disassociated from the characteristics of its loading
- How to achieve the goal?
 - Quantitative analysis
 - In-sample testing
 - Out-of-sample testing
 - Qualitative analysis



- In-sample analysis
 - Factors that have high explanatory power $R^2 = 1 \sum \varepsilon_i^2 / \sum r_i^2$
 - Issues with overfitting
 - Statistically significant: t-test, p-values, F-test $t_{score} = \hat{L}/SE_{\hat{L}}$
 - Factors should be systematic in nature, well populated
 - Residuals should be uncorrelated with each other and with systematic factors

$$\rho(\varepsilon_i, \varepsilon_j) = 0 \quad \forall i, j \qquad \rho(\varepsilon_i, F_k) = 0 \quad \forall i, k$$

- Statistical factor selection procedures
 - Discrete search
 - Continuous search



- Discrete search: Subset selection
 - + Directly attacks the problem, produces more interpretable results
 - Involves heavy combinatorial search; results are very sensitive to inputs
- Continuous search: OLS with constraints (e.g., ridge regression)

$$\hat{F} = \underset{F}{\operatorname{arg\,min}} \sum_{i} \left(r_i - \sum_{j} L_{ij} * F_j \right)^2 \qquad s.t. \quad \sum_{j} F_j^2 \le \lambda$$

- + Computationally feasible, produces more stable results
- Only shrinks the coefficients to zero, requires further analysis for selection
- Lasso estimates: A hybrid methodology

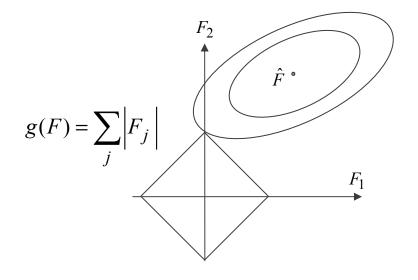
$$\hat{F} = \underset{F}{\operatorname{arg\,min}} \sum_{i} \left(r_i - \sum_{j} L_{ij} * F_j \right)^2 \qquad s.t. \quad \sum_{j} \left| F_j \right| \le \lambda$$

- Sets some coefficients to zero (controlled by λ) and shrinks the others
- Enjoys the benefits of both discrete and continuous search

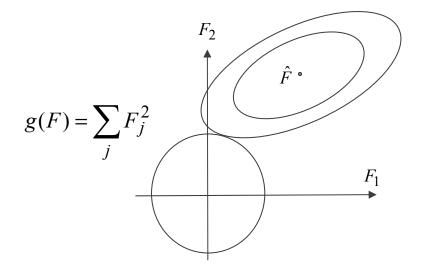


Comparing lasso and ridge regression in the case of two factors





Ridge Regression



Lasso selects corner solutions

$$F_1 = 0$$
 $F_2 = \lambda$

- Stepwise variable selection is a popular heuristic for subset selection
 - At each step of the procedure
 - Find the factor with the largest additional explanatory power
 - Check if the factor is significant in the model
 - Drop any factor if it becomes insignificant
 - Stop when no new factor can be added with the desired significance
 - US equities, % of months each fundamental / technical factor is selected

All Data		High VIX		Low VIX		First Half		Second Half	
SIZE	69%	SIZE	67%	SIZE	54%	SIZE	62%	SIZE	58%
MOMENTUM	54%	MOMENTUM	55%	MOMENTUM	49%	MOMENTUM	48%	MOMENTUM	54%
TURNOVER	53%	REALIZED VOL	52%	TURNOVER	39%	REALIZED VOL	44%	DEFAULT PROB	47%
REALIZED VOL	45%	DEFAULT PROB	50%	REALIZED VOL	36%	TURNOVER	41%	REALIZED VOL	42%
DEFAULT PROB	29%	TURNOVER	43%	DEFAULT PROB	31%	DEFAULT PROB	31%	TURNOVER	38%
TOTAL YIELD	27%	B/P	30%	TOTAL YIELD	23%	B/P	28%	FORWARD E/P	30%
FORWARD E/P	27%	FORWARD E/P	29%	B/P	20%	TOTAL YIELD	23%	E/P	24%
E/P	25%	E/P	27%	E/P	19%	E/P	21%	TOTAL YIELD	24%
B/P	23%	TOTAL YIELD	26%	FORWARD E/P	15%	ACCRUALS	10%	B/P	22%
ACCRUALS	20%	ACCRUALS	17%	ACCRUALS	12%	FORWARD E/P	8%	ACCRUALS	17%

Consistent performance of the top five factors across different sub-periods



- Prevent excess correlations: Multicollinearity
 - Why is it a problem?
 - Instable coefficient estimates, wide confidence intervals
 - How to eliminate?
 - Remove some factors or combine factors into risk indices
 - Residualize one factor to the other
- Multicollinearity also brings interpretation issues

 - $r = L * F + \varepsilon$ where $L = [L_1 \ L_2 \ . \ L_n]$ Univariate regression: The case of one factor $\hat{F}_1 = \frac{L_1^T r}{L_1^T L_1}$
 - Multivariate regression

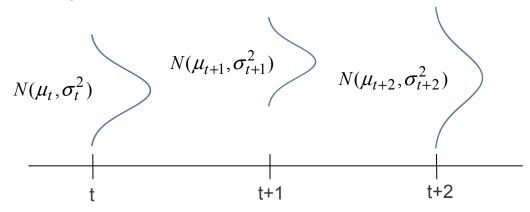
 \[\tilde{-T} = \tilde{-T} = T \]

$$\hat{F} = (L^T L)^{-1} L^T r$$

- If loadings (regressors) are uncorrelated, then L^TL is diagonal and the estimated factors have the following form L^TL
- Multivariate factors can keep $F_i = \frac{E_l}{L_i^T L_i}$ a univariate interpretation



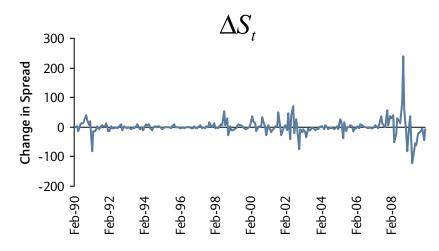
- Out-of-sample analysis: Parsimony is the key
 - Testing the accuracy of portfolio volatility estimates in an out-of-sample setting
 - Need factors that facilitate the estimation of the properties of future multivariate distribution of factors
 - Invariance: Relatively stable distribution of factor returns across time; thus, history can be used to predict future distribution



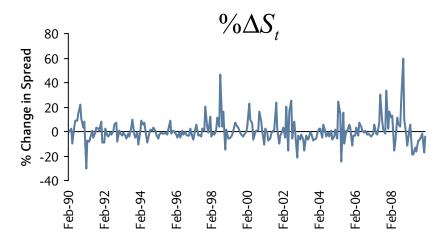
- Stability of interactions between factors: Correlations
- Need loadings that are forward looking and responsive
 - · Also need some stability to avoid excessive rebalancing



- Suppose we identify change in credit spreads as a source of returns for corporate bonds. How can we operationalize this source? $R = f(\Delta S)$
- Monthly changes in spread for the Barclays Capital USD Credit BAA Index



Statistics	ΔS_t	$\%\Delta S_t$
Skewness	2.34	1.63
Kurtosis	28.64	7.44
5% Percentile	(1.10)	(1.25)
95% Percentile	1.33	1.66

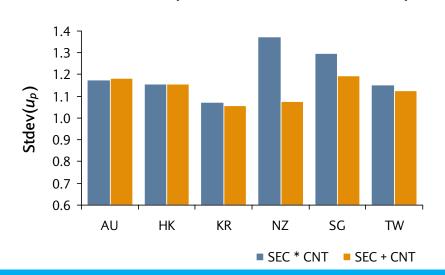


The evidence seems to suggest that $\%\Delta S_t$ is somewhat better behaved

Factor	Loading	
ΔS_t	OASD	More Stable Loading
ΔS_t	OASD*S	Conditional Loading



- Ratio test: The goal is to gauge how well the model predicts the volatility of portfolios
 - Standardize realized return with its estimated volatility: u_p^t
 - If the forecast is accurate, standard deviation of $u_{\,p}$ should be close to 1
 - Distribution of u_p is important; if far from normal, then need a large sample
 - Two competing models for equity country portfolios in ex-JP Asia
 - Sector for each country (10*6 factors)
 - Better in-sample fit, worse performance out-of-sample
 - Regional sector and country factors (10+6 factors)
 - Worse in-sample, better out-of-sample



$$u_p^t = \frac{r_p^t}{\sigma_{p,forecast}^t}$$



Factor / Model Selection – Qualitative Measures

- Intuitive factors and loadings that have clean interpretation
 - Economic sense, easy to understand and communicate
 - Statistical factors are harder to interpret
- Ability to capture major investment themes in the specific asset class
 - Interest rates, industry, fundamentals
- Target investor base
 - Short- vs. long-horizon investors
 - Market dynamics are more important for short horizon
 - Macroeconomic variables are more relevant for long horizons
 - Local investors vs. global investors
 - Local models: Good fit for local portfolios, need consistency for effective combination
 - Global models: More parsimonious, more stable estimates, may underestimate diversification benefits



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Appendix



Linear Factor Models – Estimation

- Single-step
 - Allows for a natural decomposition of total variance
 - Loss of interpretation in jointly estimated factors / loadings
 - Multicollinearity issues elevated
- Multi-step
 - Need to take a stand on the order of factors: Important factors come first
 - Order of factors makes a significant difference in their interpretation
 - First set has clean interpretation (can be univariate)
 - Others are excess / residual factors
 - Provides the setup to combine different types of factors
 - E.g., observed and estimated factors in the credit model

$$r_{BOND} = r_{carry} + \sum_{i=1}^{N} KRD_i^*(-\Delta KR_i) + Vega^*F^{\sigma} + OASD^*F^{OASD} + \varepsilon$$



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