
Using AFT and FSI Tools to Validate Components of Interest Rate Risk Models



Agenda

■ **Validation Tests of Prepayment Models**

- Models applicable to securities
- Models applicable to portfolio loans

■ **Validation Tests of Stochastic Rate Generators using FSI**

- Lattice
- Monte Carlo



Regressor and Back-testing

- *Regressor* is applicable to AFT prepayment models:
 - Securities
 - MBS
 - CMOs
- MBS data at the CUSIP level require extraction routines from Bloomberg or Intex.
- CMO data at the CUSIP level require license with Intex, Chasen or other cash flow model vendor.
- Agency CMO data can be extracted from Bloomberg.



Regressor and **Back-testing**

- **Batch “Back-test” in the *Regressor* Model**
 - **Output created:**
 - Actual CPR
 - AFT forecast of CPR using actual rates
 - **Output is written to text files with number of files equal to the number of collateral types.**
 - Example: 300 CUSIPs of 15 and 30 FNMA, FHMLC, and Whole loans might reside in six text files
 - **All CUSIPs can be run at once in *Regressor* with the text files created automatically.**
 - **Text files can be opened in EXCEL and standard “goodness of fit” or other statistics applied.**



Regressor and Back-testing

- Once a macro is written, process takes less than an hour for the entire list of CUSIPs (We did this for 300 CUSIPs.)
- Procedure can be learned in about 15 minutes via phone support, if all required extract files are available.
- *Regressor* allows the aggregation across CUSIPs drawn from same collateral. Performance on sub-portfolios can be assessed.

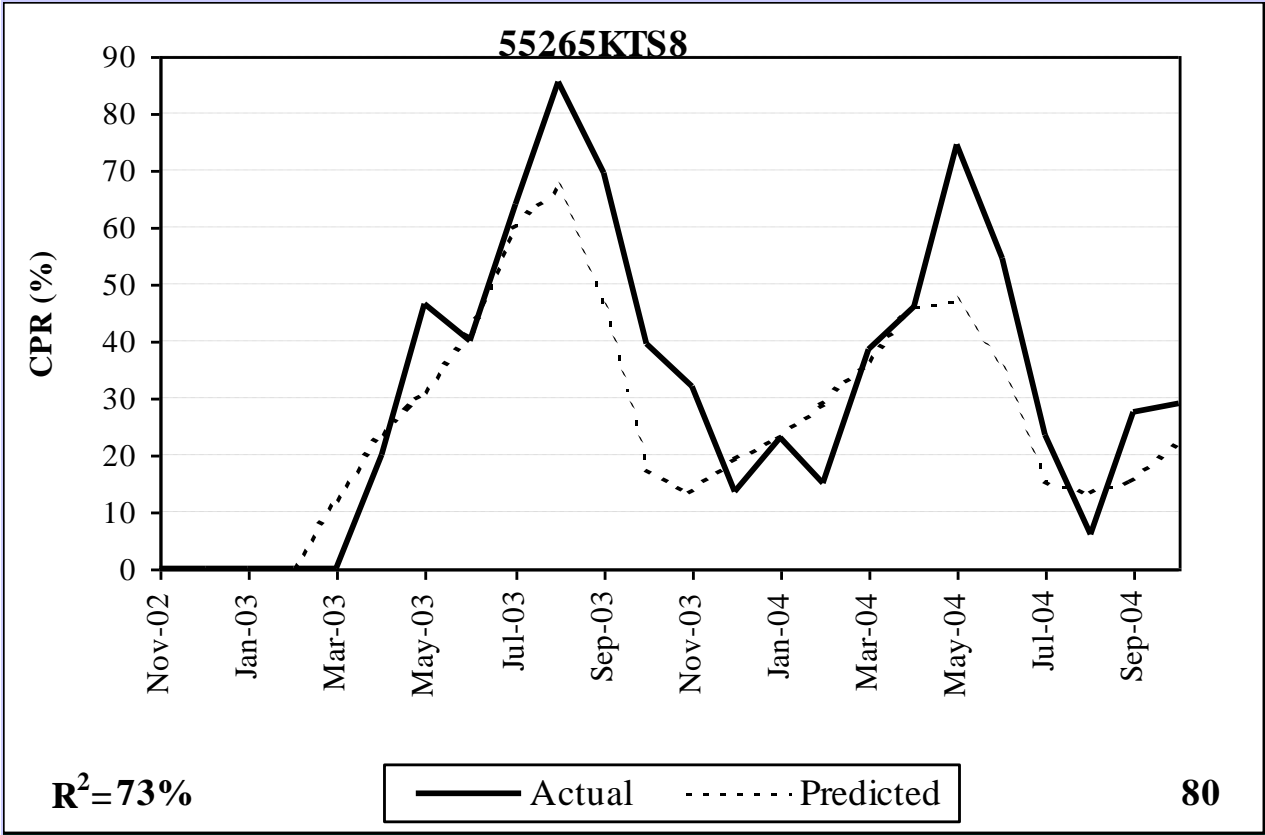


Why can AFT Claim *Regressor* Provides a Back-test”?

- AFT Prepayment models are estimated using collateral data
 - Securities are small samples of the collateral.
 - Output in *Regressor* at CUSIP level is based on both overlapping historical periods and projected periods, depending on the length of history and when collateral model was estimated.
 - Over the historical data model period this is an “in-sample” back-test.
 - When CPRs are projected beyond historical data period used for building models, the test is an “out-of-sample” back-test.
- If the connectivity in the ALM system and AFT models has been validated, then the ALM model is not needed for back-testing AFT prepayment models.



Back testing Sample Output from *Regressor*



One of 180 graphs created for an ALCO Partners client

***Dynamic Aggregator* and Back testing**

- *Dynamic Aggregator* is applicable to portfolio loans
 - Similar output created as in *Regressor*
 - AFT prepayment models provide the basis for modeling. Discrete prepayment factors can be scaled to improve model fit.
 - Historical data used to fit model and model statistics can be created as part of the estimation process.
 - In-sample and out-of-sample model testing can be applied with similar statistics.
 - If AFT-ALM model link has been validated, then the process can substitute for back test using ALM model.



Findings from *Dynamic Aggregator* Projects

- *Dynamic Aggregator* has coincidentally “solved” a classic data bottleneck in many banking institutions
 - IT departments now regularly archive historical data relevant for testing prepayment models
 - They are very efficient at providing raw data.
 - They are inefficient at “mapping” data
 - Internal resources for “research and analysis” are frequently low on a priority list
 - External projects where only raw data is required appear not to face the same internal hurdles because they demand very few IT resources



Findings from *Dynamic Aggregator* Projects

- A part of prepayment model validation projects applicable to portfolio mortgages is obtaining and mapping data. But this process is no longer a large cost.
 - Mapping costs have not been large. Key variable is number of systems
 - Mapping rules typically remain constant over time, so annual updates become very cost effective

- Once the data has been mapped into *Dynamic Aggregator*, the process for testing and is very cost-effective, particularly when the process is done annually as required by banking regulations



Testing Stochastic Rate Generators

- In many banks stochastic rate generators are being utilized to measure value based IRR
 - Base case is calibrated to market instruments
 - There are known nuances and problems associated with this
 - How does a bank “validate” a stochastic rate model?



Testing Stochastic Rate Generators

■ Approach I

- Obtain model documentation
- Build a separate model independently
- Perform double-blind tests and compare inputs and outputs

■ Issues with Approach I

- Vendors may be unwilling to provide proprietary code documentation
- Building a separate model can be very expensive, including its validation
- Specialized coding/math expertise required to “build” a stochastic rate generator



Testing Stochastic Rate Generators

- Approach II
 - Obtain a qualified stochastic rate generation system
 - Calibrate to the same set of market instruments
 - Compare inputs and outputs for tested samples

- FSI: A qualified stochastic rate generation system



Testing Stochastic Rate Generators

- FSI is a fully specified two factor lattice & Monte-Carlo system designed as a portfolio management product with validation capabilities
 - Provides a basis for comfort or discomfort depending on the model results and how the model results are integrated with portfolio management decisions
 - Since it contains both single factor (BK) and two-factor processes, it provides a basis for understanding the sensitivity of the risk measure to the underlying process utilized
 - Helps improve calibration procedures and allows a look into the “black box” of stochastic rate generation capability



Testing Stochastic Rate Generators

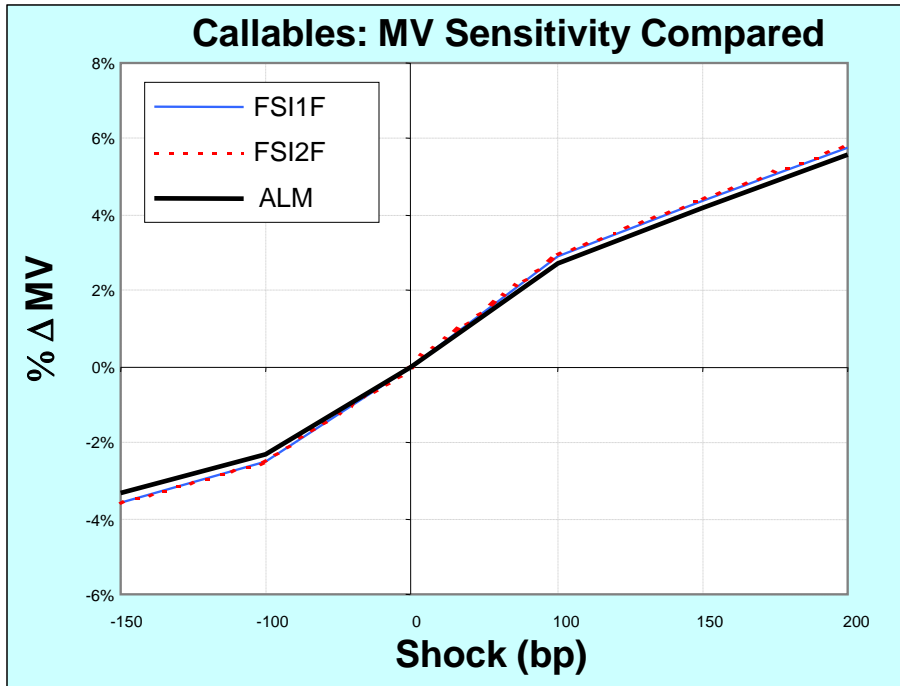
■ Test Procedures

- Compared stochastic output from a widely used ALM model with a stochastic generator
 - Compared three portfolios in two periods
 - MBS & CMO
 - Portfolio FRMs
 - Callable Bonds
 - Greeks are computed at the instrument level
 - Market values and market values under rate shocks can be computed
 - Market values changes compared over time at instrument level
 - Results used to refine and understand calibration process and reasonableness of risk results

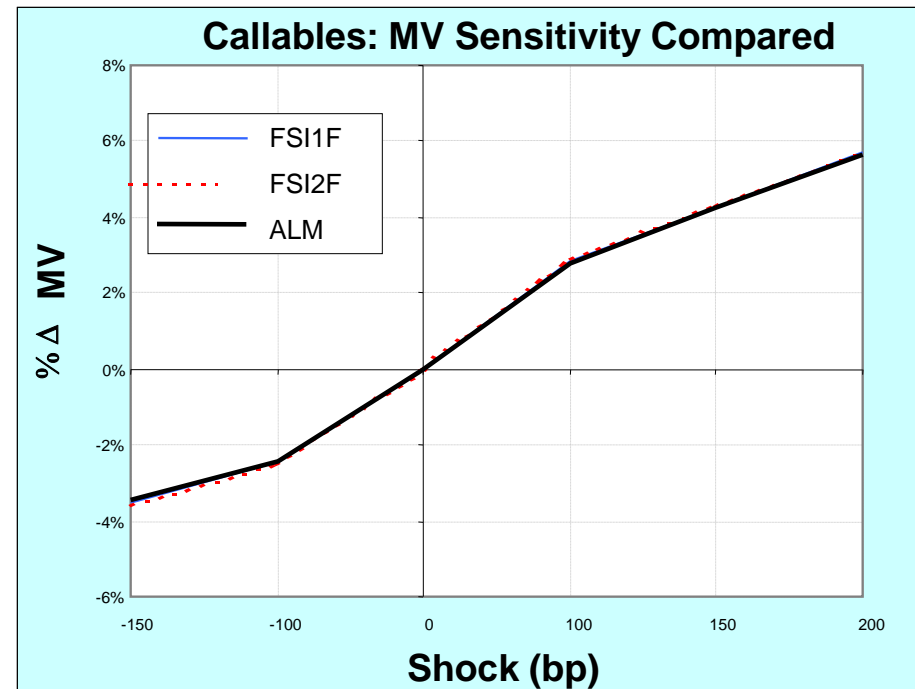


Callables: MV Sensitivity

12/31



3/30



Market value sensitivity measures for callable bond portfolio were very similar in both time periods

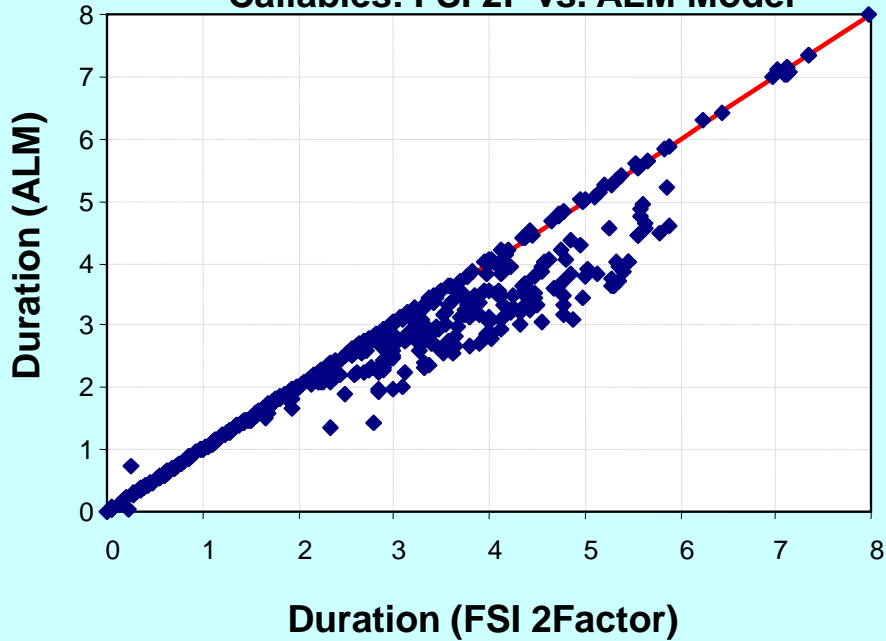


Callables: Duration & Convexity

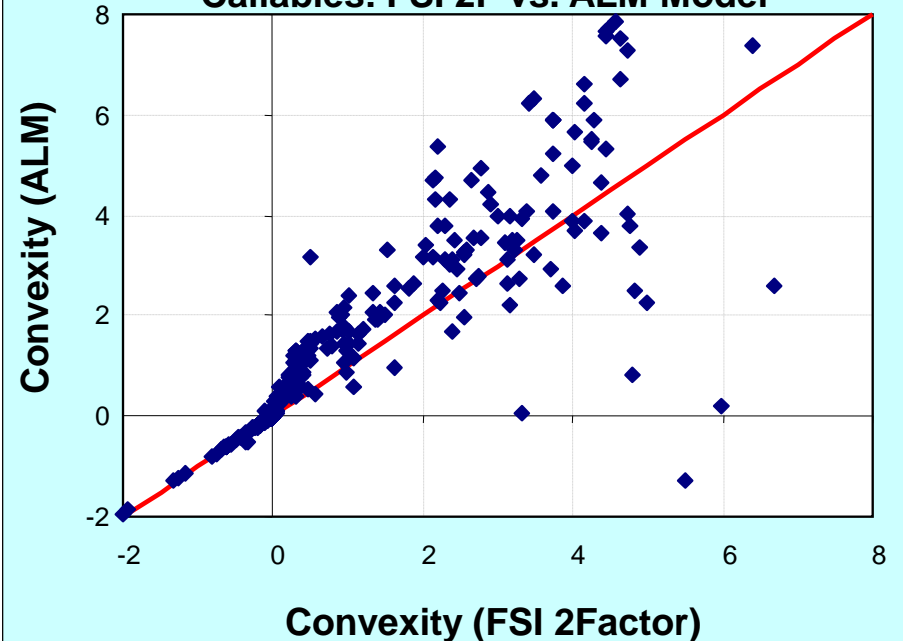
Duration

Convexity

Callables: FSI 2F vs. ALM Model



Callables: FSI 2F vs. ALM Model

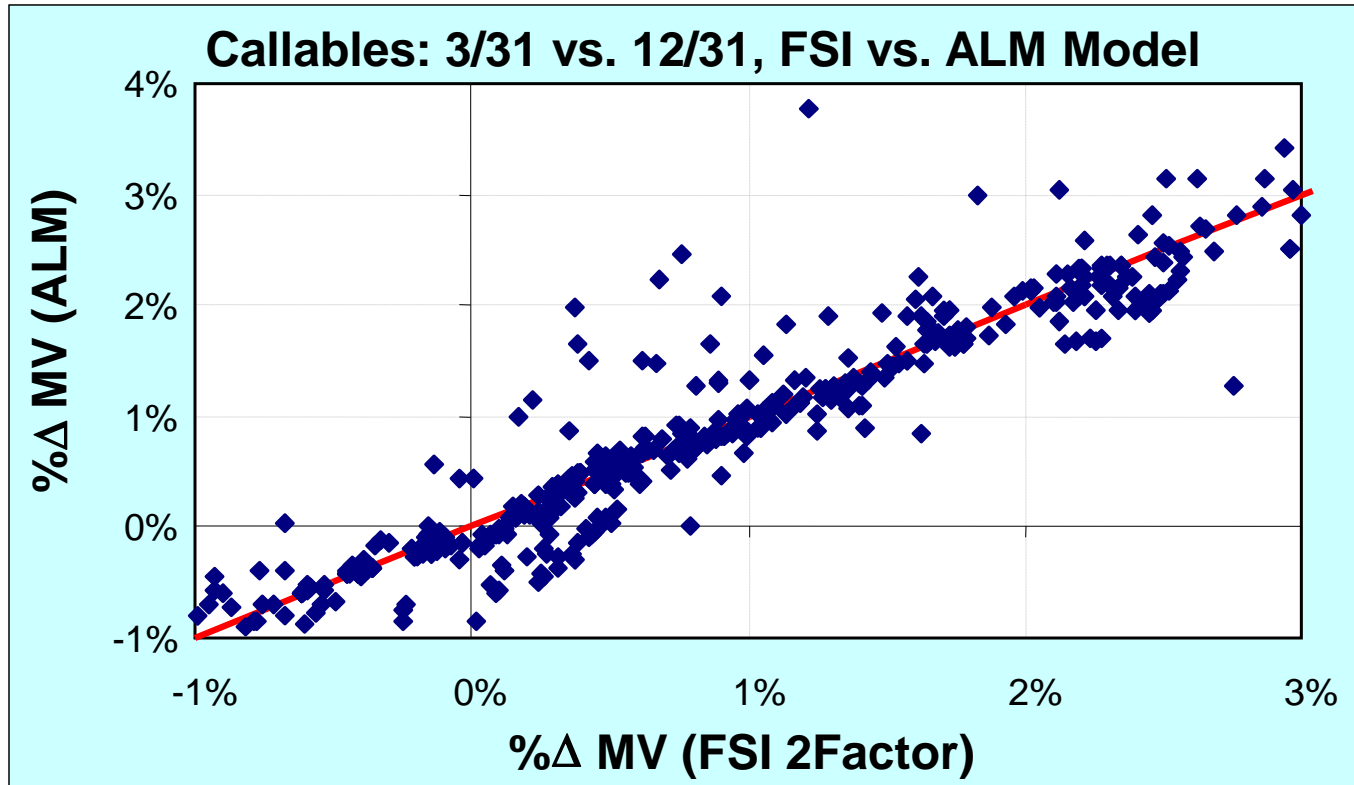


Red line represents parity

Results indicated some non-random differences on individual securities. Further investigation revealed pattern associated with “at the money” calls. Cross checks with market values showed sensitivity to calibration mechanics

Callables: *FSI 2 Factor vs. ALM Model: 12/31 to 3/31*

How does Market Value Change over Time?

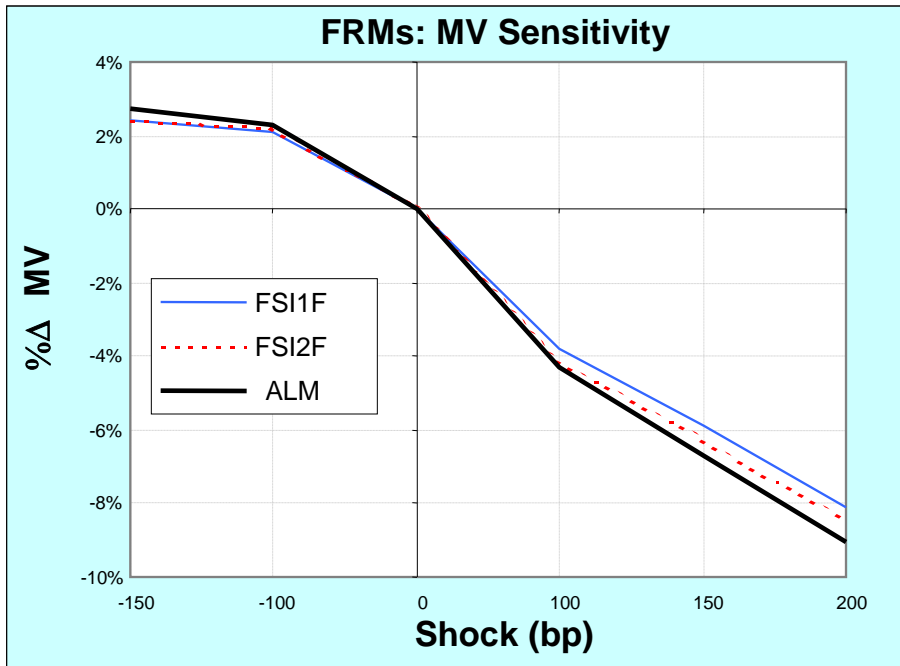


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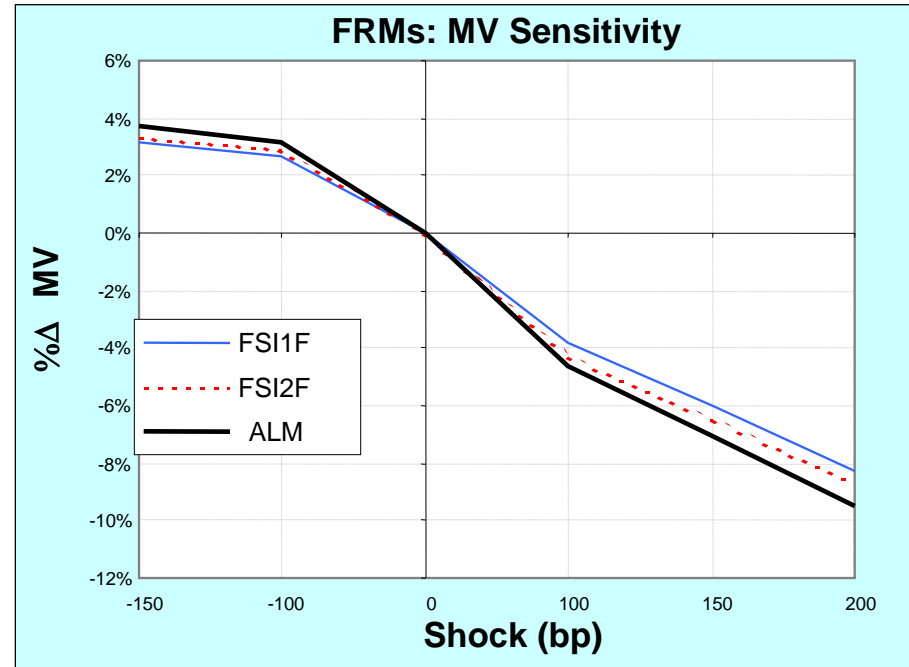


FRMs: MV Sensitivity

12/31



3/31



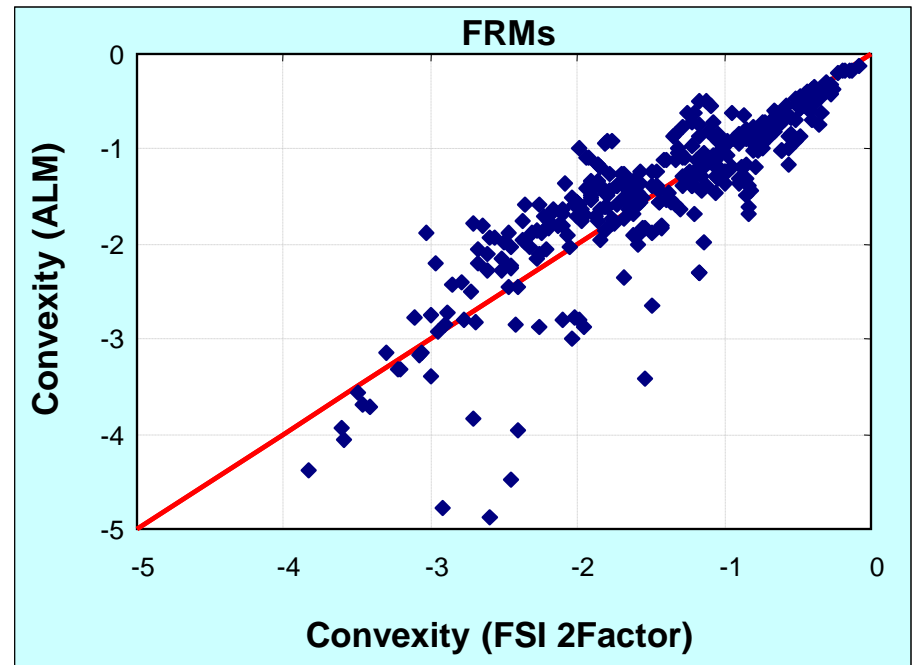
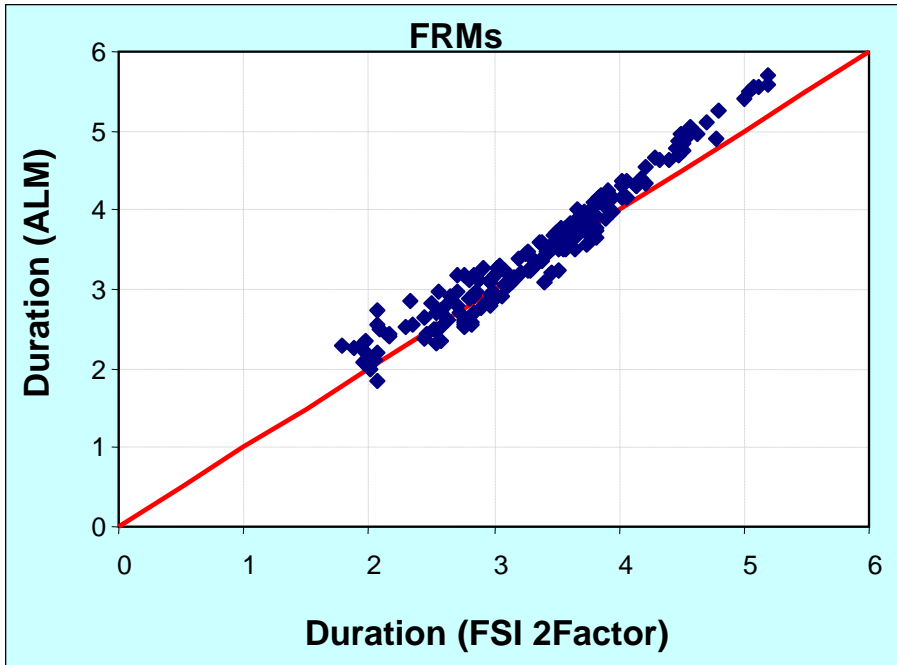
Prepayment models were identical



FRMs: Duration & Convexity

Duration

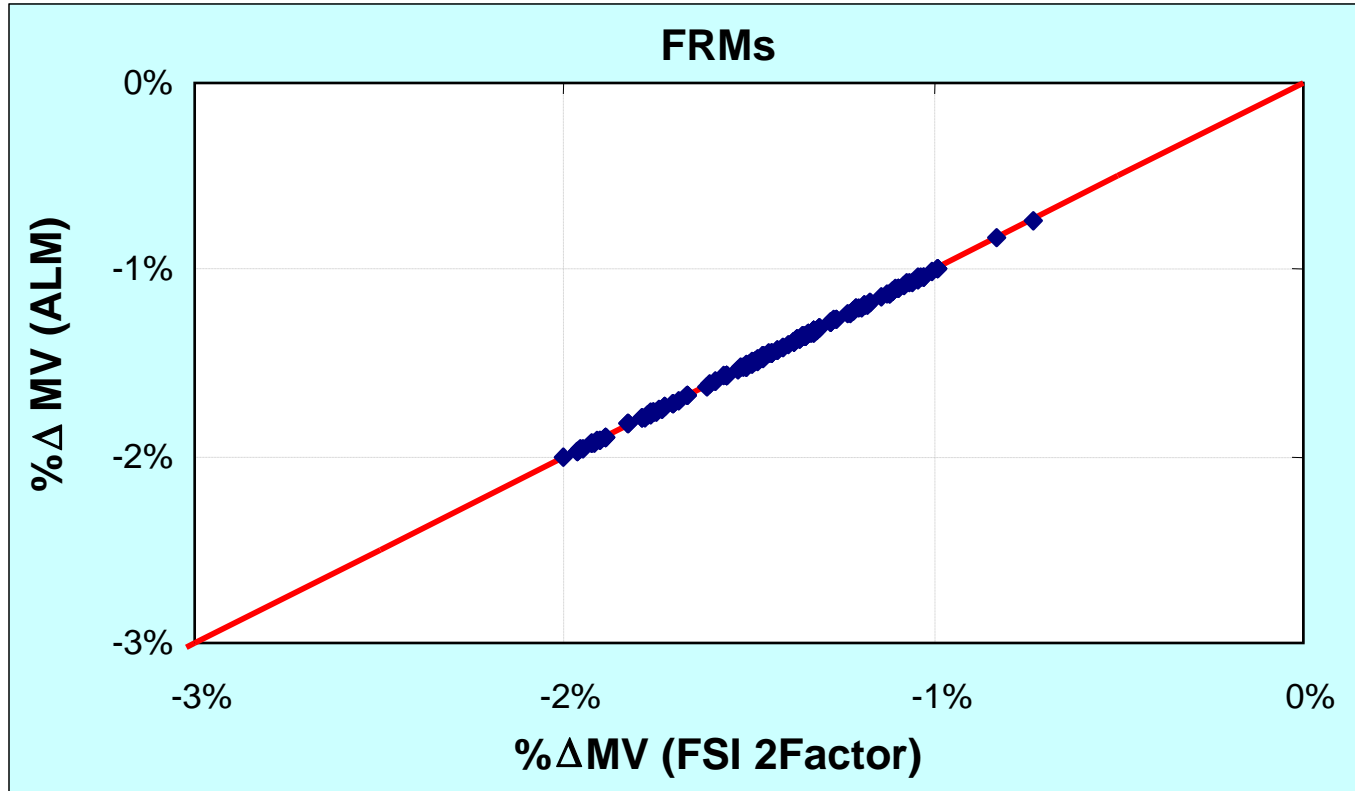
Convexity



Red line represents parity



FRMs: FSI (2 Factor) vs. ALM: 12/31 to 3/31



Red line represents parity

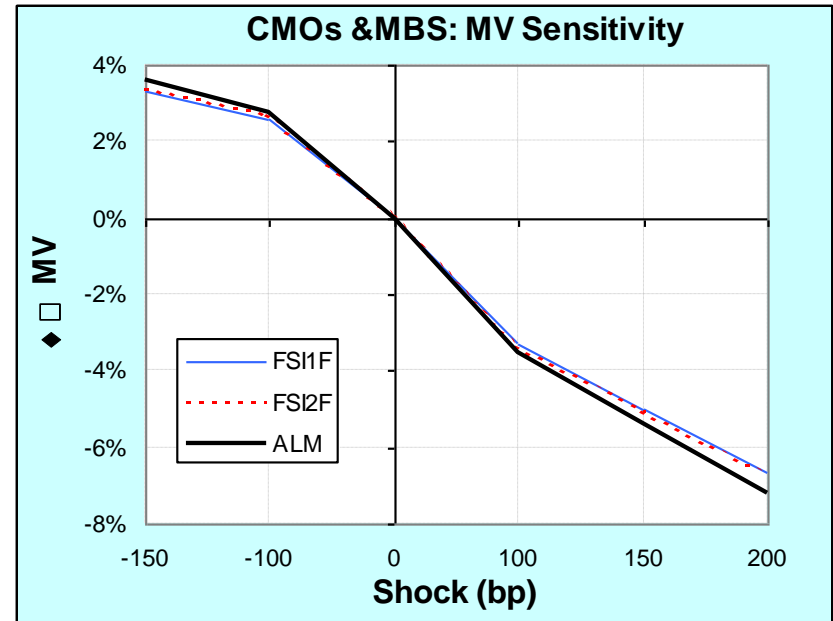
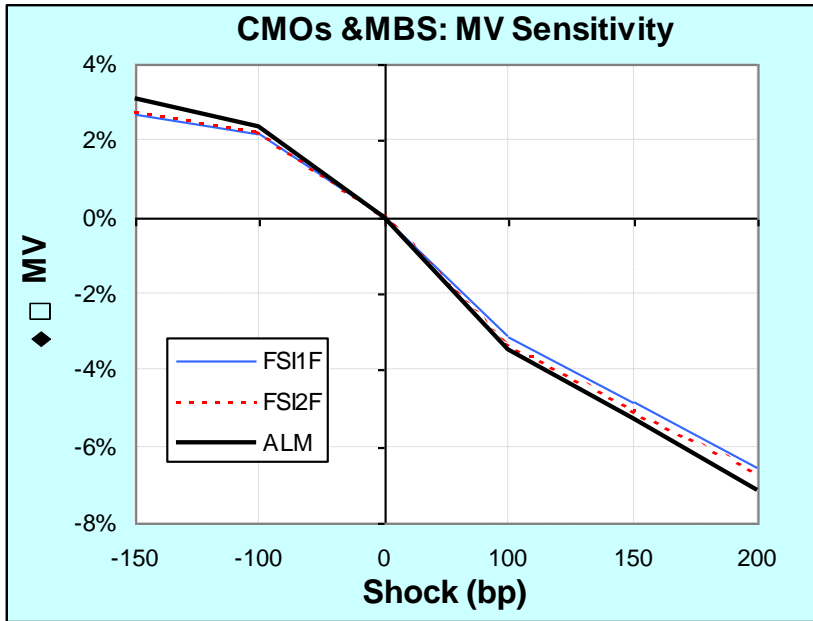
FRMs where there are good market value indicators combined with the OAS calculations lead to similar results



CMOs & MBS: MV Sensitivity

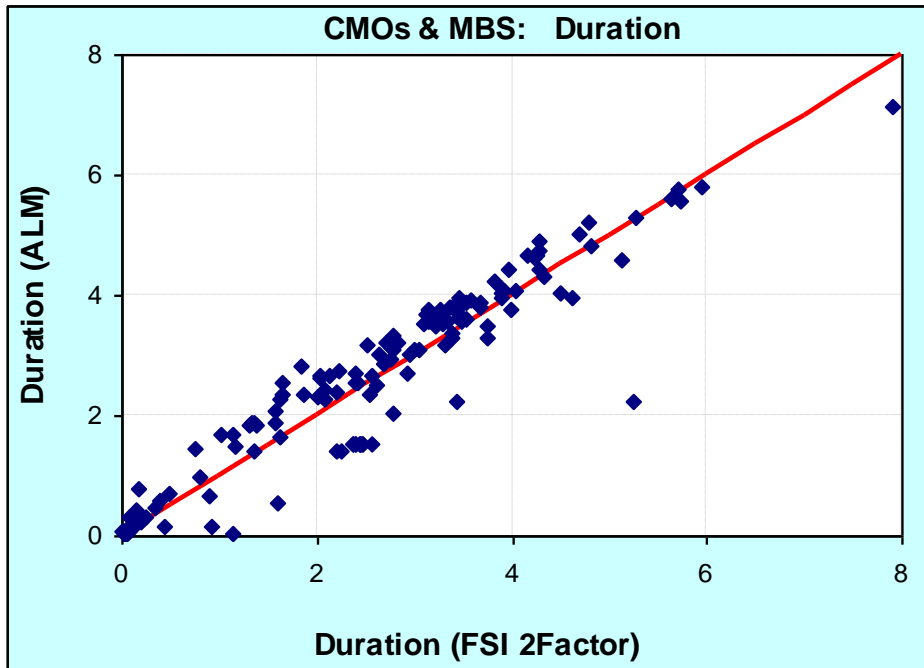
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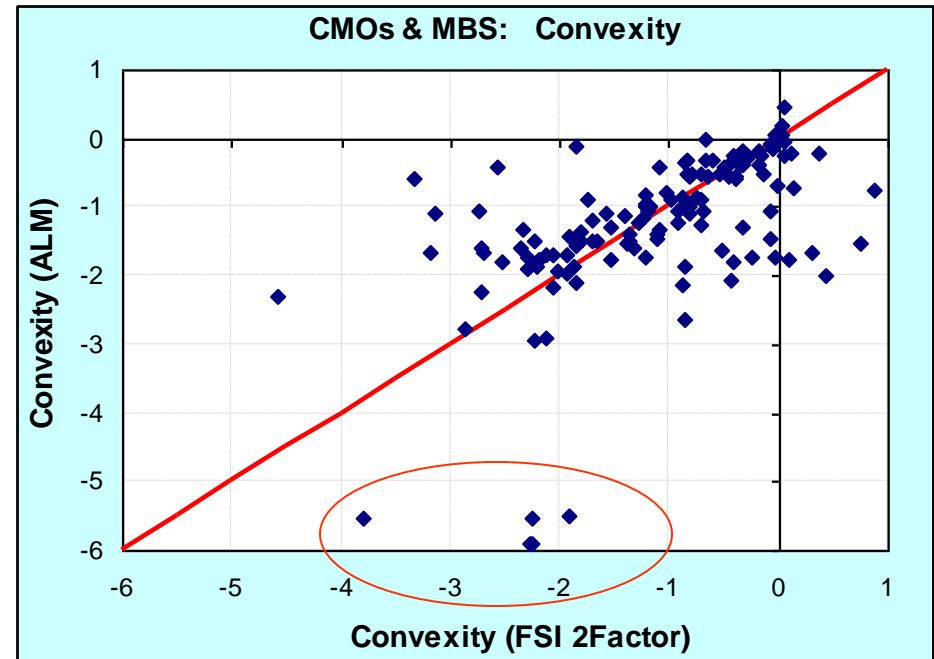


CMOs & MBS: Duration & Convexity

Duration



Convexity

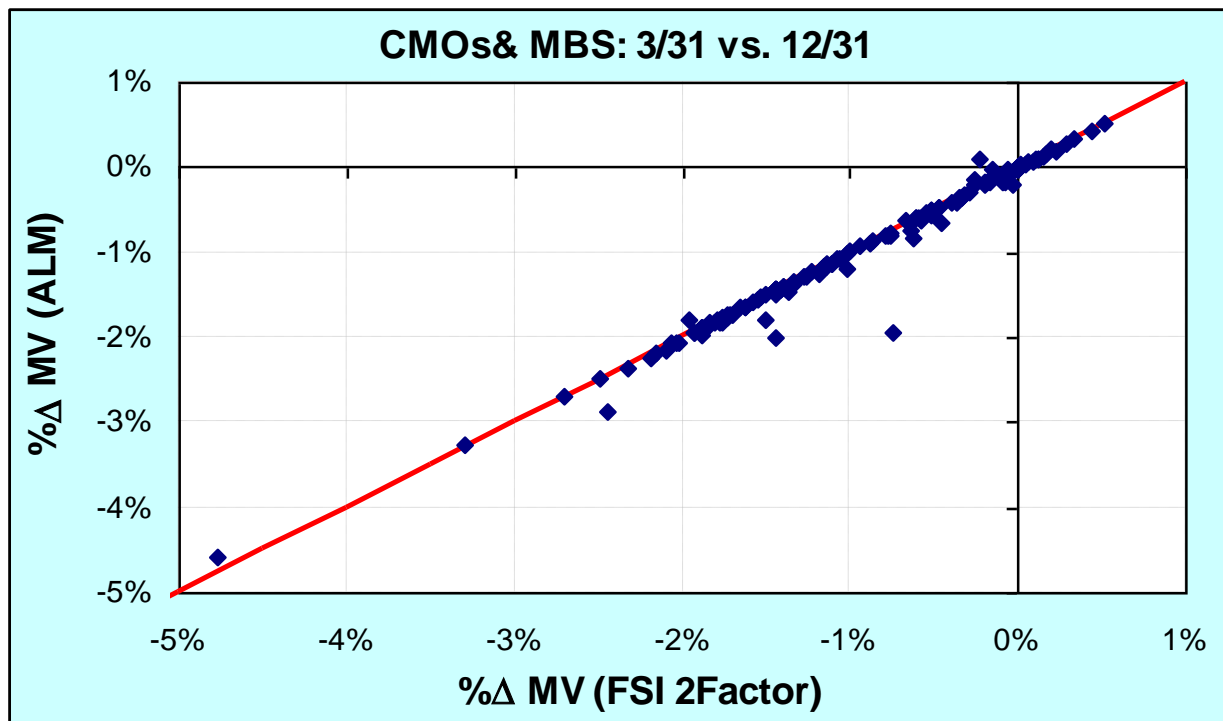


Red line represents parity

Circled outliers were later identified as input errors in the ALM model



CMOs & MBS: FSI (2 Factor) vs. ALM: 12/31 to 3/31

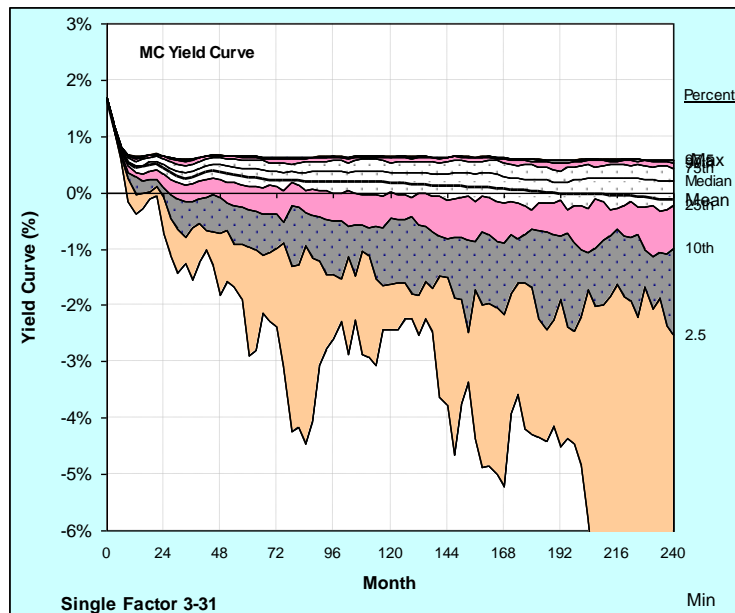
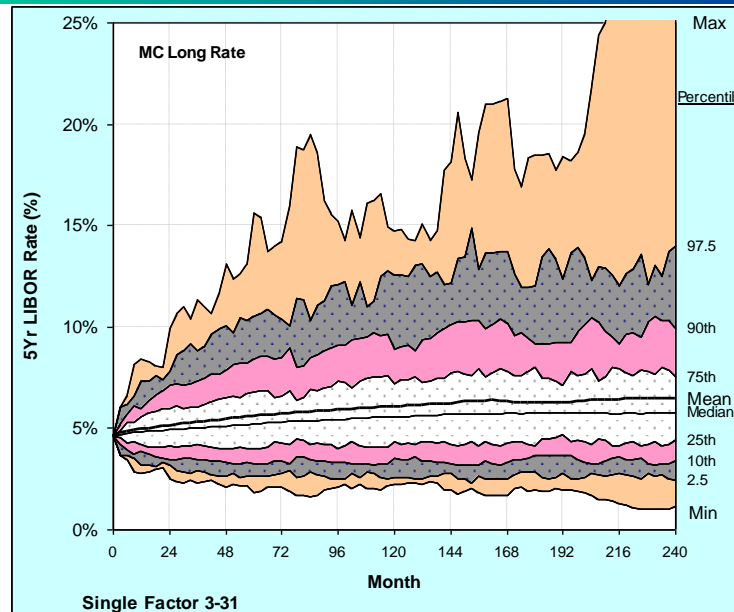
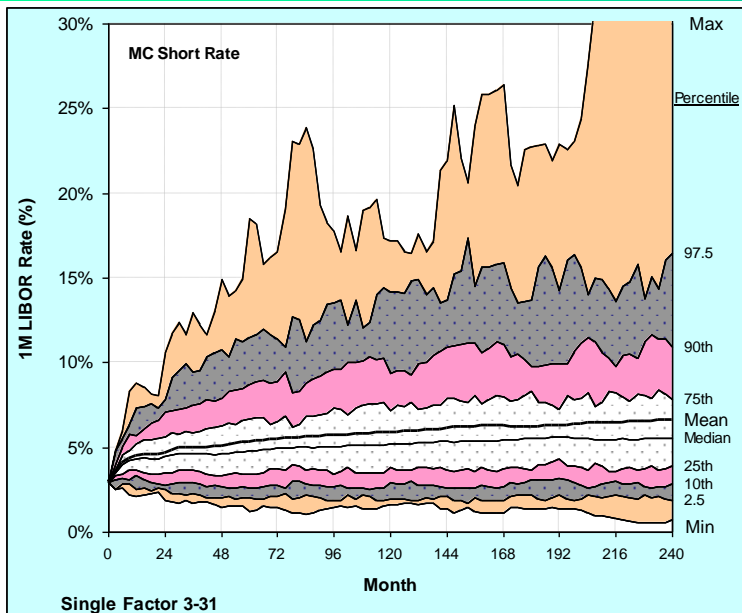


Red line represents parity

Four outliers are the same CUSIPs as in the prior graph



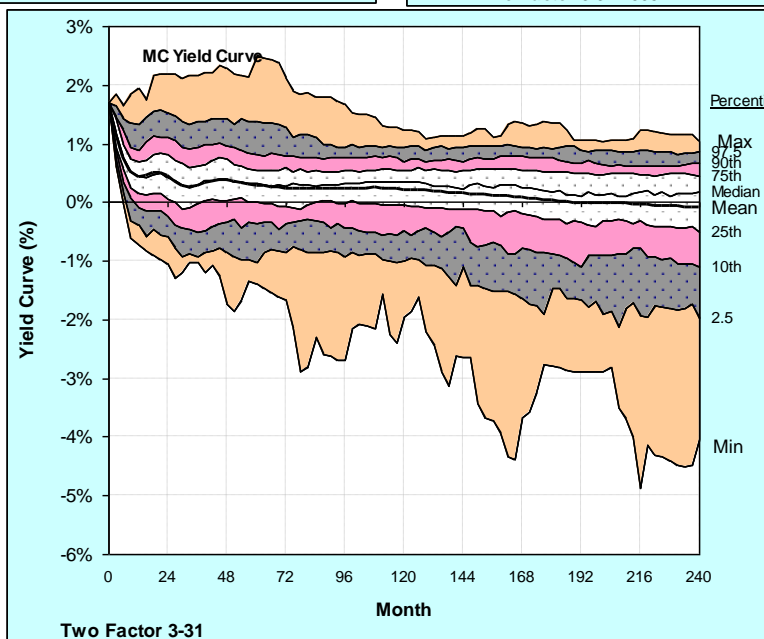
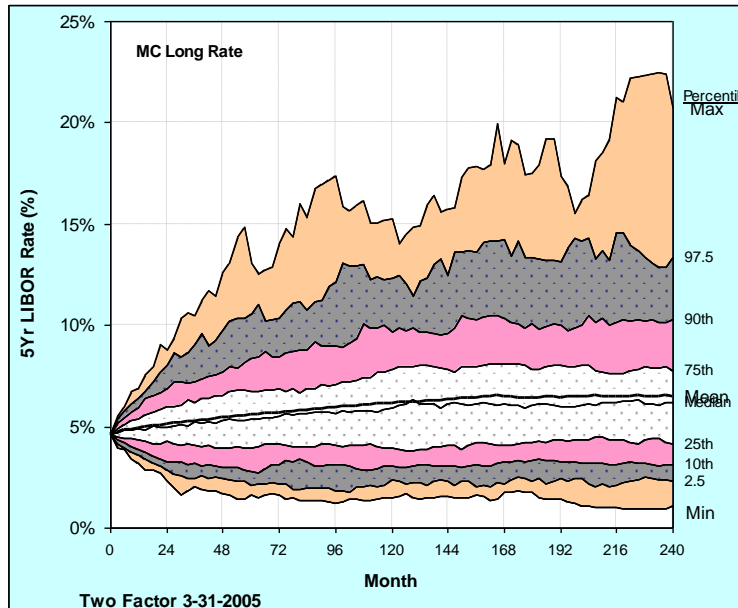
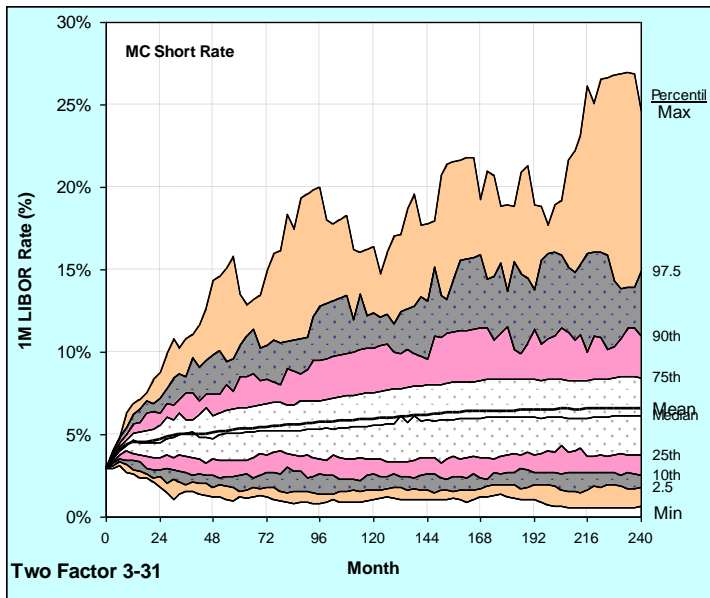
Yield Curve Distributions: Single Factor vs. Two Factor Stochastic Processes



**Single Factor Process:
Yield Curve Distribution**



Yield Curve Distributions: Single Factor vs. Two Factor Stochastic Processes

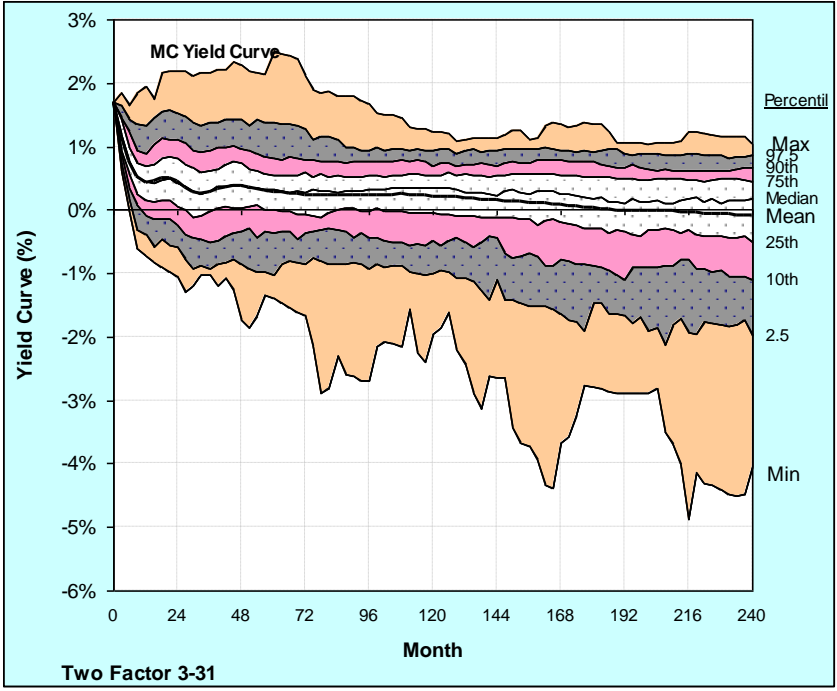
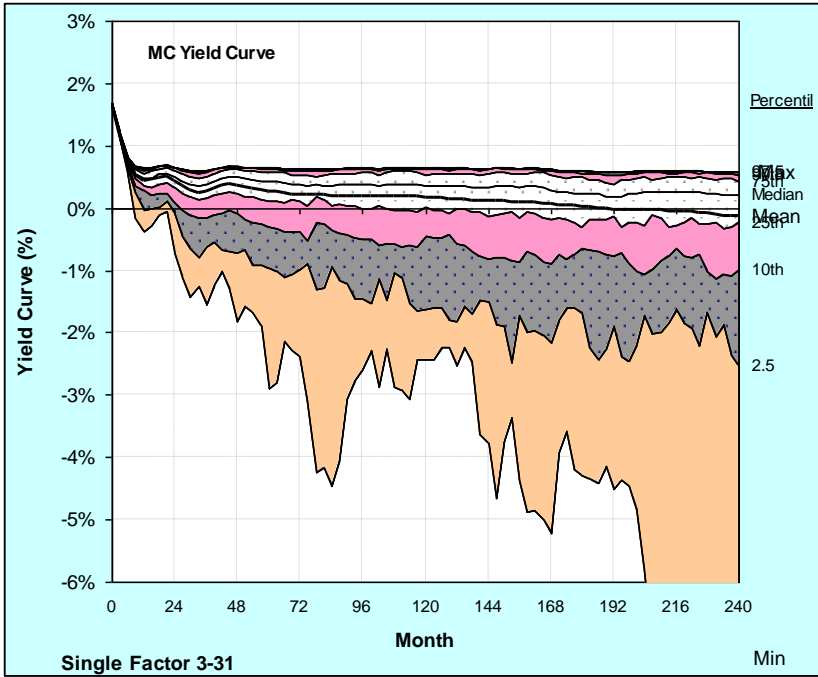


**Two Factor
Process:
Yield Curve
Distribution**

Yield Curve Distributions: Single Factor vs. Two Factor Stochastic Processes

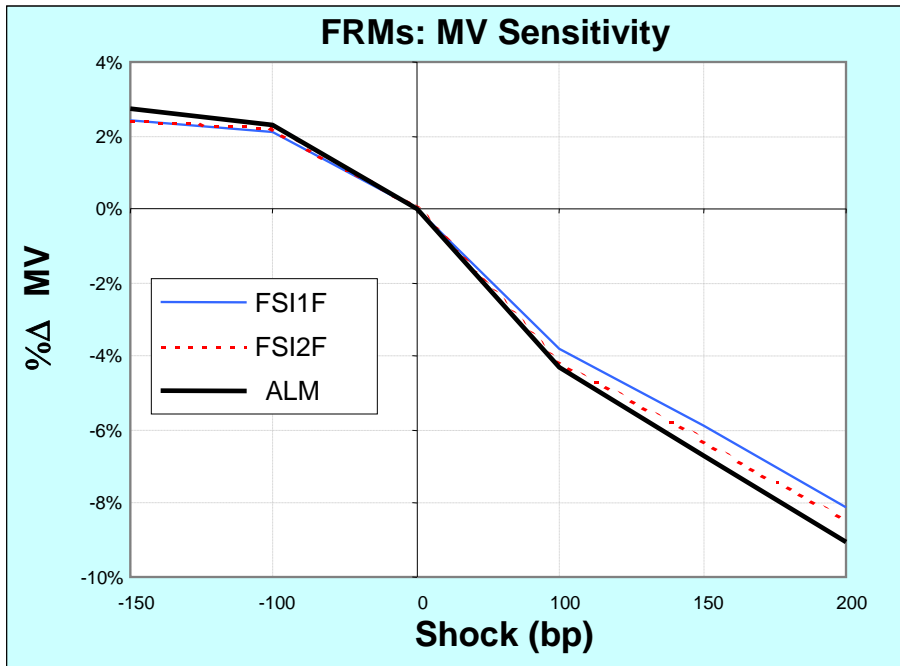
Single Factor Process:
Yield Curve Distribution

Two Factor Process:
Yield Curve Distribution

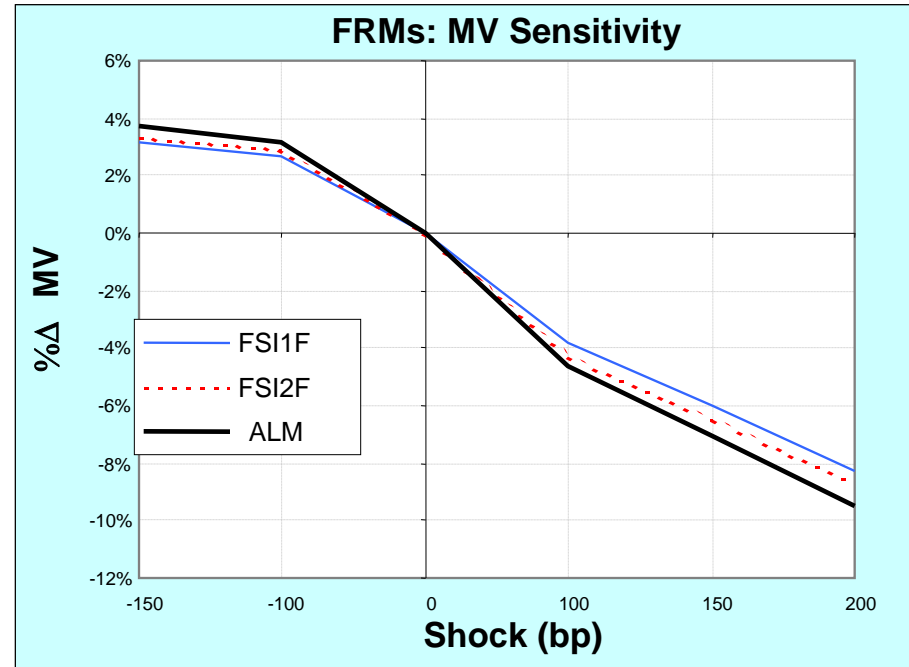


FRMs: MV Sensitivity: Single vs. 2 Factor

12/31



3/31



Impact of process is limited in this case even though distribution of yield curves is noticeable

Observations from Test

- **Stochastic models will frequently produce different measures of risk at the instrument level for instruments with embedded options**
 - We believe – but it is difficult to confirm without access to source code – that this is largely due to differences in calibration procedures as well as differences in modeling methods.
 - With larger portfolios that include instruments with a wide variety of option structures, the differences in results will tend to cancel each other out if the overall levels of implied volatility used in the calibrations are similar.



Observations from Test

- **Validation tests of stochastic models using a qualified stochastic model, such as FSI, can:**
 - Identify modeling errors, because large differences in results invite further investigation
 - Enhance understanding of the “black box” of ALM model calibration procedures, when calibrations aren’t “all that close”
 - Provide guidance for the product characteristics that are stochastic modeling technique sensitive
- **Validation tests of stochastic models lead to the following warning:**

Beware of relying on a single stochastic model when measuring risk at the instrument level for instruments with embedded complex options



Observations from Test

Tests of stochastic models can identify modeling errors, because large differences invite further investigation

Word to the warning about using stochastic models for risk management involving individual transactions

