

PIMCO Advisory's Approach to RMBS Valuation

December 8, 2010



The Building Blocks of Valuation

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National House Price Situation

As Of September 30, 2010

YoY		MoM		Peak	Peak to	Peak to
Current	Previous	Current	Previous	Month	Current	Trough
0.5%	1.6%	-0.8%	-0.5%	Apr-06	-29.6%	-31.8%
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SOURCE: S&P, Haver Analytics







Mortgage Loan Credit Model

The role of a mortgage loan credit model is to project mortgage default, prepay and loss severity on a loan basis – and by extension – on a mortgage-backed security

PIMCO Advisory's loss expectations are determined by employing proprietary loan level quantitative models

The proprietary loan-level default model has the following three major components:

- Incorporates borrower and property characteristics based on attributes known at the time of origination
- Includes dynamic performance data from origination including borrower payment, interest rates, and home price histories
- Incorporates future economic information on regional home-price appreciation and mortgage/interest rates

We employ different sub-models by credit (subprime, Alt-A etc.) and product type (fixed vs. adjustable)

The output of the model is a set of CPR, CDR and severity vectors



Loan-Level Default Analysis

PIMCO Advisory's loan level process models each individual loan in each securitization

Historical loan performance is a critical factor in projecting future performance

- In our model, loans are classified into two groups: performing and non-performing
- The model incorporates not only the current statt56(r)3.61868.33333 0 0 6(e)-2.949794856(r)3.61869(r)36 Tm [(P)-1.7



Illustrative Path-Dependent Simulation



Loss Severity Analysis

For the purposes of loss severity, the same default probabilities are applied to maintain consistency. Additional components that contribute to the ultimate loss severity analysis include:

- Collateral deficiency (unpaid balance less REO sales price)
- Lost interest (accrued as servicer advances)
- Expenses (legal, property taxes, brokerage fees)
- Mortgage insurance considerations

The explanatory variables incorporated in the default probability have a linear relationship with loss severity. Historical trends can help predict loss severity sensitivities to inputs such as:

- Static Factors (at origination)
 - FICO
 - Property type
 - Occupancy (owner, investor/second)
 - Lien-Position
 - Mortgage Insurance
 - Judicial vs. non-judicial state
- Dynamic Factors
 - Interest-rate
 - Loan Balance
 - HPA/HPD
 - Current marked-to-market LTV
 - Regional Foreclosure Timelines
 - Time/Loan Age

Sample Adjustment: Accounting For Loan Modification s





Model Curves are Input into the Cash-flow Engine an d Allocated According to Each Deal's Waterfall Structure

Group 1	Group 2						
	A2D ^K						
A1A	A2A						
A1B	A2B						
A1C	A2C						
	A2D ^K						
M1							
M2							
M3							
M4							
M5							
M6							
B1							
B2							
B3							
B4	Ļ						
B5	5						

Bonds get paid principal and interest and losses at a particular point in time

The deal's legal documents determine the waterfall rules

	6			
Group 1		Group 2		
AV1 ^{(37.25%) i}	AF1 ^{(0.00%) i}	AF2 ^{(37.25% i} M1 ^(25.67%) M2 ^(14.76%) M3 ^(8.12%) M4 ^(1.99%) M5 ^(0.00%) M6 ^(0.00%)	AF3 ^{(37.25%) i}	AF4 ^{(37.25%) ik}
		B5 ^(0.00%)%)%)	%)%7.5 525pp	