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Market Risk Management

FRM二级知识框架图



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Framework of Market Risk Management

VAR and Other Risk Measures

- Parametric And Non-parametric Methods Of Estimation
- Backtesting VAR
- VAR Mapping
- Correlation Risk Modeling and Management
- Academic Literature On Risk Measuring

Risk Management Of Fixed Income and Option

- The Science Of Term Structure Models
- LIBOR VS. OIS
- Empirical Approaches To Risk Metrics And Hedging
- Volatility Smiles



Section 1



PARAMETRIC AND NON-PARAMETRIC METHODS OF
ESTIMATION

Non-parametric Methods

basic HS approach: ★

two steps

- **Order** (sort) the daily profit/loss observations
- Locate the loss **corresponding to the specified confidence level**
- 95% VAR is the $(n \cdot 5\% + 1)$ th highest observation. → 结论与一级不同

Expected Shortfall (Conditional VAR)

- expected value of the loss when it exceeds VAR
- **average of the tail loss / average of tail VARs**
- as a subadditive risk measure
- **VAR定义模糊, ES定义清晰, 有唯一答案**

bootstrap historical simulation (数据有限)

- draw a sample, records the VAR → repeat over and over → multiple sample VARs → average of them
- Empirical analysis: provides more precise estimates than historical simulation on raw data alone

Historical Simulation
Approach

(数据权重 $1/n$)

**Weighted Historic
Simulation Approaches**
(数据权重不同)

Age-weighted HS ★
(hybrid of EWMA & HS)

$$w_{(i)} = \frac{\lambda^{i-1}(1-\lambda)}{1-\lambda^n} \rightarrow$$

计算

advantages

- Generalizes historical simulation (HS) ($\lambda \rightarrow 1$)
- more responsive to large losses & handling clusters of large losses
- reduce ghost effects
- option of letting our sample size grow over time

Volatility-weighted HS → weight returns by relative volatility

$$r_{t,i}^* = \left(\frac{\sigma_{T,i}}{\sigma_{t,i}} \right) \times r_{t,i} \rightarrow$$

advantages

- Directly accounts for volatility changes
- Allows us to incorporate GARCH forecasts
- Can obtain estimates exceed maximum loss in actual datasets
- Empirical evidence to support superiority of estimates

Correlation-weighted HS → major generalization of the volatility-weighted approach

Filtered HS

- bootstrapping + traditional HS with conditional volatility modeling (GARCH)
- Semi-Parametric Bootstrap

do not make strong assumptions about the distribution

↓ 优缺点 定性了解★

Advantages	Disadvantages
<ul style="list-style-type: none">• Intuitive• Handle non-normal returns (skew, kurtosis)• handle any position type including derivatives• Popular among risk practitioners• Relatively easy to implement (e.g., spreadsheet)• No “curse of dimensionality”• Uses readily available data (returns, volatility)• Easy to report and communicate results• Easy to generate confidence intervals• Can be combined with parametric “add-ons”	<ul style="list-style-type: none">• If history is a “quiet period,” VAR (or ES) estimates will be too low• Difficulty handling in-sample shifts (or regime changes)• Can be dominated by extreme losses (not helpful if unlikely to occur)• Most are subject to ghost effect (shadow effect)• Generally make no allowance for plausible events that might occur but did not occur• To an extent constrained by largest loss in historical dataset

Parametric Estimation Approaches

Normal VAR ★ 计算

- returns that follows a normal distribution

$$\text{VAR}(\alpha\%) = -\mu_{P/L} + \sigma_{P/L} \times z_{\alpha}$$

Lognormal VAR ★ 计算

- returns that follow a lognormal distribution

$$\text{VAR}(\alpha\%) = P_{t-1} \times (1 - e^{\mu_R - \sigma_R \times z_{\alpha}})$$

Other Issues

Coherent Risk Measures ★ ★

- Monotonicity: if $X_1 \leq X_2$, $\rho(X_1) \geq \rho(X_2)$
- Translation Invariance: $\rho(X + k) = \rho(X) - k$
- Homogeneity: $\rho(bX) = b\rho(X)$
- Subadditivity: $\rho(X_1 + X_2) \leq \rho(X_1) + \rho(X_2)$

not coherent
risk measure

coherent risk
measure

- **Standard deviation** violates
 - the monotonicity
 - translation invariance condition
- **VAR** is not
 - sub-additive
- **Spectral** measures are coherent:
 - weighted average of the quantiles
 - (所有数据点加权; 越往左, 权重越大)
- **ES** (special case)
 - (只有尾部等权重加权, 其余部分权重为0)

Standard Errors Of Coherent Risk Measures

- **sensitive to the choice of n (sample size)**: $n \uparrow$, the quantiles will be further into the tails where more extreme values of the distribution are located
- tails became heavier, **ES** estimators became **more prone to the effects of large but infrequent losses**
- presence of heavy tails might make **ES** estimators in general **less accurate** than VAR estimators

QQ plot ★

- predicted quantile against the empirical quantile
- **linear** QQ plot indicates a **good fit** to the distribution
- departures from linearity can tell whether the tails of empirical distribution are fatter, or thinner

Other Issues

★ Extreme Value Theory(EVT)

- central limit theorems do not apply to extremes
- extremes are governed by extreme-value theorems

Block Maxima Method → *Generalized Extreme Value Distribution (GEV)*

params: location(μ), scale(σ), shape/tail(ξ) + select threshold(u)

Fisher-Tippett theorem: sample size(n) \uparrow , distribution of extremes converges to the GEV distribution

- same goal and general principles of EVT
- same shape/tail parameter: ξ

tradeoff of the threshold

high enough so that the GPD applies
low enough so that there are sufficient observations

Peaks over Threshold → *Generalized Pareto Distribution(GPD)*

params: positive scale(β), shape/tail(ξ) + select threshold(u)

distribution of excess loss over u : $F_u(x) = P\{X - u \leq x | X > u\}$

extreme events are not independent

POT相比GEV

POT Advantages

POT Disadvantages

- GEV involves an additional parameter
- problem of choosing the threshold
- and involves some loss of useful data

Multivariate EVT → 相关性

ρ : 线性, normal (elliptical) distribution

copulas: 非线性, not normal



Section 2



BACKTESTING VAR

Backtesting VAR

Backtesting VAR

process of comparing losses predicted by the VAR model to those actually experienced over the testing period

Difficulties in backtesting a VAR model



statistical decision(not 100% confidence)

assumes a static portfolio

- contamination minimized in short horizons
- track both actual and hypothetical return
- cleaned-return (consider Transaction fee)

Process Of Model Verification

★ 计算

failure rate(Bernoulli trials) → binomial probability distribution(x: number of exceptions)

$$f(x) = \binom{T}{x} p^x (1-p)^{T-x}$$

假设检验: central limit theorem(binomial distribution → normal distribution)

$$z = \frac{x - pT}{\sqrt{p(1-p)T}} \sim N(0,1)$$

- H₀: accurate model
- H_a: inaccurate model

Other Issues

Type I And Type II Errors →

- **Probability of Type I error** (mistakenly reject null) = α = level of significance
- **Probability of Type II error** (mistakenly accept null) = β (difficult to calculate)

Null: Model is correctly calibrated		
Decision	Null is correct	Null is incorrect
Accept *	Good decision!	Type 2 Error
Reject	Type 1 Error	Good decision!

* Technically, “accept” = “Do not reject”
(as a jury finds defendant “not guilty” instead of “innocent”)



Unconditional Coverage Model: ignores time variation (reflects “independence” in the i.i.d. assumption)

Conditional coverage models: analyze exceptions conditional on current condition

Backtesting Exceptions

Using Failure Rates in Model Verification

- H_0 : accurate model
- H_a : inaccurate model

★ 重要数字

- Test statistic (LR)
- If $LR > 3.84$, we would reject the hypothesis (the model is correct)

The Basel Rules For Backtesting

Basel internal approach(market risk) : 10-day 99.0% VAR model ★★



Over 250 observations in one year, the expected number of exceptions is $0.01 * 250 = 2.5$

Four categories of causes for exceptions:

- The basic integrity of the model is lacking.
The **penalty** should **apply**.
- Model accuracy needs improvement.
The **penalty** should **apply**.
- Intraday trading activity.
The penalty should be **considered**.
- Bad luck. The market significantly varied.

The Basel Committee decided ★★

Three Zones		
Zone	# of Exceptions	Increase in Scaling Factor (k)
Green	0-4	0
	5	0.40
	6	0.50
Yellow	7	0.65
	8	0.75
	9	0.85
Red	10 or more	1.00



Section 3



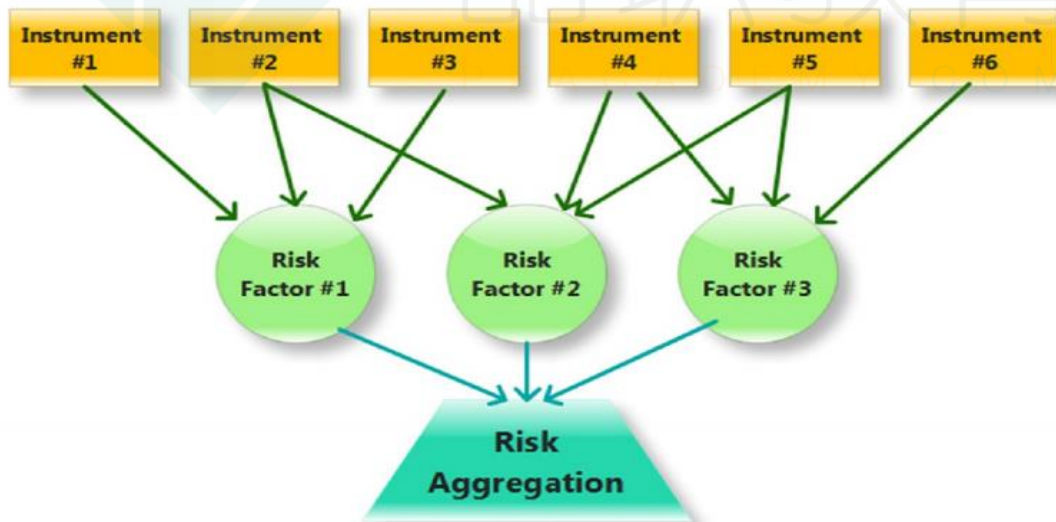
VAR MAPPING

VAR Mapping

Mapping replaces current portfolio position values with exposures to risk factors.

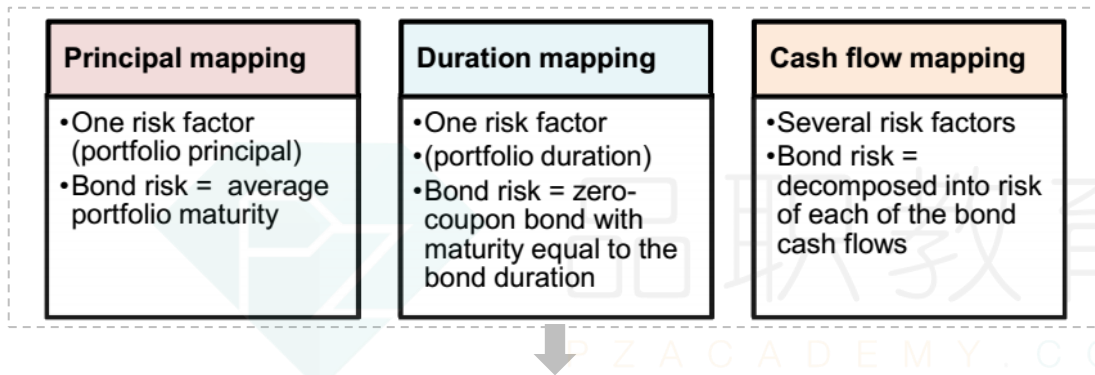
Mapping Principles

- positions can be simplified to smaller number of positions on set of risk factors
- when instrument characteristics change over time.
- only solution to data (shortage) problem



Mapping Portfolios Of Fixed Income Securities

★ Mapping Approaches to fixed-income portfolios → 掌握基础班讲义例题★



linear Derivatives → *Delta-normal method*

$$\begin{cases} \text{VAR}(dP) = |-D^* P| \times \text{VAR}(dy) \\ \text{VAR}(dc) = |\Delta| \times \text{VAR}(dS) \end{cases}$$

nonlinear Derivatives → *Delta-Gamma method*

$$\begin{cases} \text{VAR}(dP) = |-D^* P| \times \text{VAR}(dy) - (1/2)(C \times P) \times \text{VAR}(dy)^2 \\ \text{VAR}(dc) = |\Delta| \times \text{VAR}(dS) - (1/2) \Gamma \times \text{VAR}(dS)^2 \end{cases}$$

★记忆方法: 凸性的好处→降低风险

Other Issues

Stress Testing

- This stress test generates a distribution of present value factors that can be used to re-price the portfolio

VAR Relative To A Benchmark Portfolio

- Benchmarking a portfolio: measuring VAR relative to a benchmark portfolio
- Tracking Error VAR = $\alpha \sqrt{(x - x_o)' \sum (x - x_o)}$



Section 4



CORRELATION RISK MODELING AND MANAGEMENT

Some Correlation Basics

通常假设 ρ 不变 $\xrightarrow{\text{correlation change}}$ Financial Correlation Risk

the risk of financial loss due to adverse movements in correlation between two or more (financial) variables

Eg. Wrong-way risk: adverse correlation between a bond issuer and the bond insurer

危机时的 ρ 变化规律

影响 ρ 因素 nonfinancial variables

During a severe systemic **crisis**, an **increase in correlation** is typical

High diversification is related to low correlation

- Sovereign debt and currency value
- Geopolitical tensions
- Correlated markets and economies

Trading And Correlation

hedge

Multi-Asset Options

correlation \downarrow , option price \uparrow , except Option on the worse of two

- **Option on the better of two.** Payoff = $\max(S1, S2)$.
- **Option on the worse of two.** Payoff = $\min(S1, S2)$.
- **Call on the maximum of two.** Payoff = $\max[0, \max(S1, S2) - K]$.
- **Exchange option (as a convertible bond).** Payoff = $\max(0, S2 - S1)$.
- **Spread call option.** Payoff = $\max[0, (S2 - S1) - K]$.
- **Option on the better of two or cash.** Payoff = $\max(S1, S2, \text{cash})$.
- **Dual-strike call option.** Payoff = $\max(0, S1 - K1, S2 - K2)$.

Quanto option(protects an investor against currency risk)

know two things

- how deep in the money
- exchange rate

Tools to Hedge Correlation Risk

直接hedge ρ

Correlation Swap ★ : *fixed correlation* \leftrightarrow *realized/stochastic correlation* →

payer → long correlation → correlation ↑ gain

receiver → short correlation → correlation ↓ gain

$$\rho_{\text{realized}} = \frac{\sum_{i>j} \rho^{i,j}}{n \times (n-1) / 2}$$



间接hedge σ

→ *positive relationship between correlation and volatility*

Buy Call On Index And Sell Call On Individual

- Long Index Call
- Short Component Call

Variance Swap

- pay fixed in a variance swap on an index and
- to receive fixed in variance swaps on individual components of the index.

If correlation increases, so will the variance. ★

Correlation And Financial Crisis Of 2007 To 2009

了解

Causes Of The Crisis

- Increased speculation fueled by loose credit
- Complex credit products
- Model risk
- Moral hazards
- Increased risk appetite

First correlation-related crisis

- May 2005
- the correlations of the assets in the CDO decreased, the hedge funds lost on both positions

Correlation Risk and Risk Management

Market risk (comprised of equity risk, interest rate risk, currency risk, and commodity risk)

Market risk implicitly incorporates correlation risk

结论

Credit risk 构成

- Migration risk
- Default risk

the higher the correlation, the higher the probability of total loss

事实

period 1981 to 2001

- Default correlations between industries are mostly positive with the exception of the energy sector
- The default correlation within sectors is higher than between sectors

sector-diversified loan portfolio to reduce default correlation risk

term structure of defaults

- most investment grade bonds, the term structure of default probabilities increases in time
- bonds in distress, the default term structure is typically inverse

Systemic risk ➡ Systemic risk and correlation risk are highly dependent

Concentration risk ➡ High sector diversification reduces default risk

How Do Correlations Behave in the Real World

结论

Equity Correlation Behaviors

- Correlation levels are lowest in strong economic growth times.
- In recessions, correlation levels typically increase.
- Correlation volatility is lowest in an economic expansion and highest in worse economic states.
- Positive relationship between correlation level and correlation volatility

Mean reversion

- Fixed coupon bonds exhibit strong mean reversion
- Interest rates are assumed to be mean reverting

经济周期

Autocorrelation

Time lag 2 autocorrelation is highest★

the autocorrelation of 22.49% and the strong mean reversion of 77.51% sum to 1.0

Statistical Correlation Models

limitation of financial model →

- random models better replicate human behavior
- need to be stress-tested
- limitations were ignored in the crisis of 2007 to 2009

线性

parametric →

Pearson Correlation →

$$\rho(X, Y) = \frac{\text{Cov}(X, Y)}{\sigma(X)\sigma(Y)}$$

5 Limitation

- Linear dependencies
- Zero correlation derived in equations does not mean independence
- natural dependence measures only if the joint distribution is elliptical
- variances of the sets X and Y have to be finite
- not invariant to transformations

nonparametric
(ordinal measure)

Spearman's Rank Correlation

$$\rho_S = 1 - \frac{6 \sum_{i=1}^n d_i^2}{n(n^2 - 1)}$$

Kendall's τ

$$\tau = \frac{n_c - n_d}{n(n-1)/2}$$

- concordant pair
- discordant data pair
- pair is neither concordant nor discordant

Problems ★

- ordinal correlation are less sensitive to outliers
- Kendall τ : many non-concordant and many non-discordant pairs may occur

Copula Functions

非线性

Purpose Of Copula Functions ➔

- simplify statistical problems
- multiple univariate distributions → single multivariate distribution
- n-dimensional function → unit-dimensional one

Gaussian Copula(correlation) ➔

most applied copulas in finance

$$C_g[G_1(u_1), \dots, G_n(u_n)] = M_n[N^{-1}(G_1(u_1)), \dots, N^{-1}(G_n(u_n)); \rho_M]$$

P Z A C A D E M Y . C O M



Section 5



ACADEMIC LITERATURE ON RISK MEASURING

Academic Literature on Risk Measuring

- no unique solution
- square-root-of-time (“square root rule”) → inaccurate approximation

Time Horizon

proper time horizon → 取决于 portfolio liquidity and purpose for risk measurement
incorporate time-varying volatility into VAR models → 否则会 **underestimation of risk**
accuracy of simple VAR measures diminish as time horizon lengthens

Backtesting VAR

less effective over longer time horizons
conditional backtests improve power of backtests

Liquidity Risk

Exogenous liquidity → transaction costs → “liquidity-adjusted VAR” approach
Endogenous liquidity → price impact → depends on trade size (move market prices)

→ **Endogenous liquidity effects are important when**

- underlying asset is not very liquid
- size of the position
- small investors follow the same hedging strategy
- asymmetric information → magnifies the sensitivity of prices to clusters of similar trade

Practical Issues

Compare Risk Measures Methods

- Value at Risk (VAR) → not subadditive
- Expected Shortfall → subadditive
- Spectral risk measures → generalization of expected shortfall

Leverage

- procyclical
- inversely related to the market value of total assets
- cyclical feedback loop
- amplify boom and bust cycles

Within bank's risk assessment framework

- **compartmentalized approach**: sum measured risks separately
- **unified approach**: consider the interaction among various risk factors
- calculating individual risks and adding them together is not an accurate measure of true risk due to risk diversification

The Basel approach is a non-integrated approach to risk measurement

Top-down vs. Bottom-up

- **top-down approach**: risks are separable and can be aggregated in some way
- **bottom-up approach**: better account for the interaction among risk factors



Section 6



THE SCIENCE OF TERM STRUCTURE MODELS

Bond Valuation

Discounted Future CF →
$$P = \sum_{t=1}^n \frac{\text{cash flow}_t}{(1 + \text{discount rate})^t}$$



Interest rate risk → Duration: bond's interest rate risk or sensitivity of a bond's full price to a change in its yield

Macaulay Duration: the average time to wait for each payment, weighted by PV of cash flow

★
$$\text{Macaulay duration} = \sum_{t=1}^n [t \times (PVCF_t / P_0)]$$

Yield-Based DV01 →
$$\text{DV01} = \text{Duration} \times \text{Bond Value} \times 0.0001 = D^* \times P \times 0.0001$$
 →
$$DV01 = -\frac{1}{10,000} \times \frac{dP}{dy}$$

Yield-Based Convexity →

- Convexity is always positive for regular coupon-paying bonds
- Greater convexity is beneficial both for falling and rising yields

→ **All else equal, duration and convexity both increase for** ★

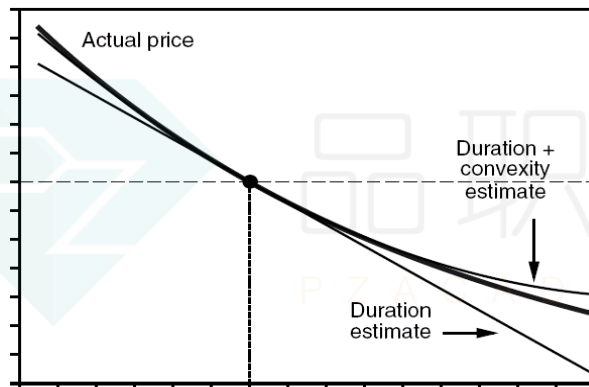
- longer maturities(夜长梦多)
- lower coupons(财大气粗)
- lower yields(看斜率)

Bond Valuation

Bond Price and Yield →

•Using Duration only: $\Delta P = -D^* \times P \times \Delta y$

•Using Duration and Convexity: $\Delta P = -D^* \times P \times \Delta y + \frac{1}{2} \times C \times P \times (\Delta y)^2$



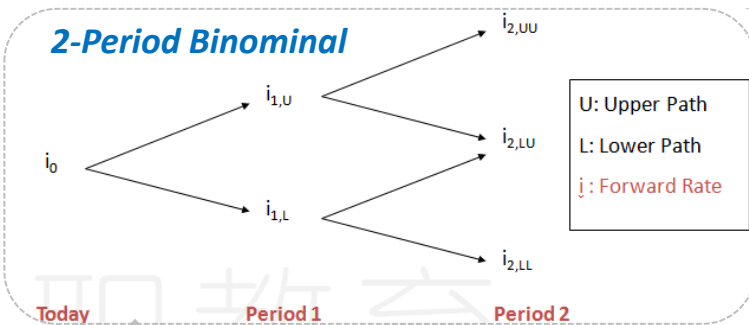
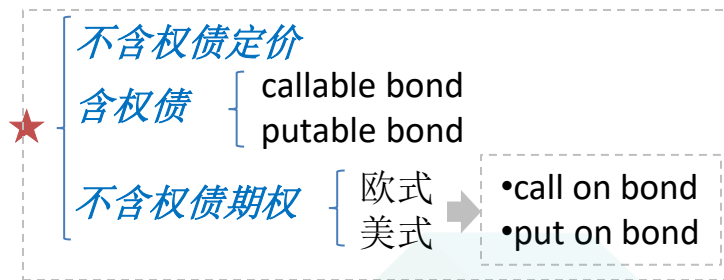
The Barbell vs. The Bullet →

•**barbell bond portfolio**: short maturities (low duration) + long maturities (high duration)

•**bullet bond portfolio**: concentrates in medium-term maturities

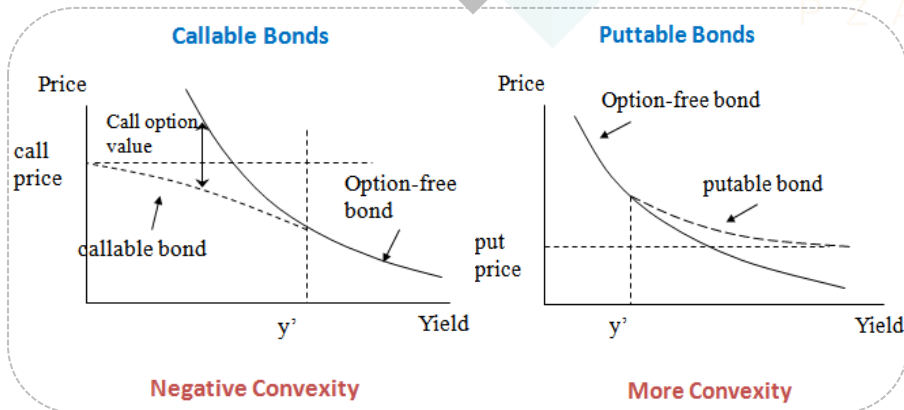
With the same duration, barbell bond has larger convexity than bullet bond

The Science of Term Structure Models



Interest Rate Tree (Binominal) Model: to value bonds with embedded options

No Arbitrage Rule



Black-Scholes-Merton model not appropriate to value derivatives on fixed-income securities. It assumes:

- no upper limit to the price of the underlying asset. However, bond prices do have a maximum value.
- risk-free rate is constant. However, changes in short-term rates do occur.
- volatility is constant. However, price volatility decreases as bond approaches maturity.

The Shape of The Term Structure

Theory	Details
Pure Expectation	forward rates are solely a function of expected future spot rates 不足: fail to consider the riskiness of bond investing
Liquidity Preference	forward rates are biased estimates of the market's expectation of future rates include liquidity premium
Market Segmentation	<ul style="list-style-type: none">• shape of the yield curve → preferences of borrowers and lenders• Yield at each maturity is determined independently
Preferred habitat	<ul style="list-style-type: none">• forward rates represent expected future spot rates plus a premium(not related to maturity)• explain almost any yield curve shape

Interest Rate Volatility: volatility of expected rates causes the future spot rates to be lower

★ Convexity Effect(Jensen's inequality) $\Rightarrow E\left[\frac{1}{(1+r)}\right] > \frac{1}{E[1+r]}$

the value of convexity increases with maturity and volatility

The Art of Term Structure Models: Drift

★
计算

Model 1 (no Drift) $dr = \varepsilon\sigma\sqrt{dt}$

Trend **Model 2 (with Drift)** $dr = \lambda dt + \varepsilon\sigma\sqrt{dt}$

Ho-Lee Model (with time-dependent drift) $dr = \lambda(t)dt + \varepsilon\sigma\sqrt{dt}$

**mean
reverting**

→ **Vasicek Model**

$$dr = k(\theta - r)dt + \varepsilon\sigma\sqrt{dt}$$

k=a parameter that measures the speed of reversion adjustment
 θ =long-run value of the short-term rate assuming risk neutrality
 r =current interest rate level

Assuming a long-run interest rate of r_L , then long-run mean reverting level is: $\theta \approx r_L + \frac{\lambda}{k}$

Model 3 (with time-dependent volatility) $dr = \lambda(t) + \varepsilon\sigma e^{-\alpha t}\sqrt{dt}$

Cox-Ingersoll-Ross (CIR) Model $dr = k(\theta - r)dt + \sigma\sqrt{r}\varepsilon\sqrt{dt}$ **σ与r相关**

Model 4 (Lognormal) $dr = ardt + \sigma r\varepsilon\sqrt{dt}$

↓ **结论★**

$$d[\ln(r)] = a(t)dt + \sigma\varepsilon\sqrt{dt}$$

$$d[\ln(r)] = k(t)[\ln\theta(t) - \ln(r)]dt + \sigma(t)\varepsilon\sqrt{dt}$$

- Ho-Lee model, the drift terms are **additive**
- Lognormal model, the drift terms are **multiplicative**



Section 7



LIBOR VS. OIS

LIBOR VS. OIS

衍生品定价要用到 r_f

常见 r_f

Treasury bills

→ artificially low

原因

- regulatory requirements
- capital of a bank, required
- favorable tax treatment

LIBOR

→ not totally risk-free

→ 3-month LIBOR-OIS spread

- normal market conditions: 10 bps
- October 2008: high of 364 bps

overnight indexed swap (OIS) rate

应用

- non-collateralized derivatives → LIBOR
- collateralized derivatives → OIS rate

- a swap: fixed rate → geometric average of overnight rates
- OIS rate: fixed rate in an OIS
- OIS: short lives
- OIS longer than 1 year is divided into 3-month subperiods
- OIS rate: continually refreshed overnight rate
- might be a default on an overnight loan
- might be a default on the OIS swap itself

- The OIS zero curve is as close to risk-free.
- It should be used for discounting regardless of whether the transaction is collateralized

r_f



Section 8



EMPIRICAL APPROACHES TO RISK METRICS AND
HEDGING

Empirical Approaches To Risk Metrics And Hedging

Drawback with a DV01-neutral →

- “parallel” assumption → but “twists” in the curve → “curve risk”
- not necessarily a realistic hedge

improve →

- empirical data → not co-dependent
- Advantage: estimate how much nominal yield changes given a change in TIPS yield
- Advantage: estimate of the volatility of the hedged portfolio

Regression hedge →

$$\text{Portfolio} = \text{Bond} + \text{Hedge} \Rightarrow \begin{cases} \Delta P = \Delta B + \Delta H = BD_B \Delta y_B + HD_H \Delta y_H = 0 \\ \Delta y_B = \beta \Delta y_H \end{cases}$$

正负号表示long/short

★计算

Two variable regression →

$$P = B_{20} + H_{10} + H_{30} \Rightarrow \begin{cases} \Delta P = \Delta B_{20} + \Delta H_{10} + \Delta H_{30} \\ = B_{20} D_{20} \Delta y_{20} + H_{10} D_{10} \Delta y_{10} + H_{30} D_{30} \Delta y_{30} = 0 \\ \Delta y_{20} = \beta_{10} \Delta y_{10} + \beta_{30} \Delta y_{30} \end{cases}$$

Principal Component Analysis(PCA) →

- sum of the variances of the principal components equals the sum of the variances of the individual rates
- principal components are uncorrelated with one another
- principal components are chosen to have the maximum possible variance



Section 9



VOLATILITY SMILES

What Is Volatility Smiles

Volatility smile : a plot of the **implied volatility** of an option as a function of its **strike price**



Put-Call Parity(whatever pricing model is used): $p + S_0 e^{-qT} = c + X e^{-rT}$



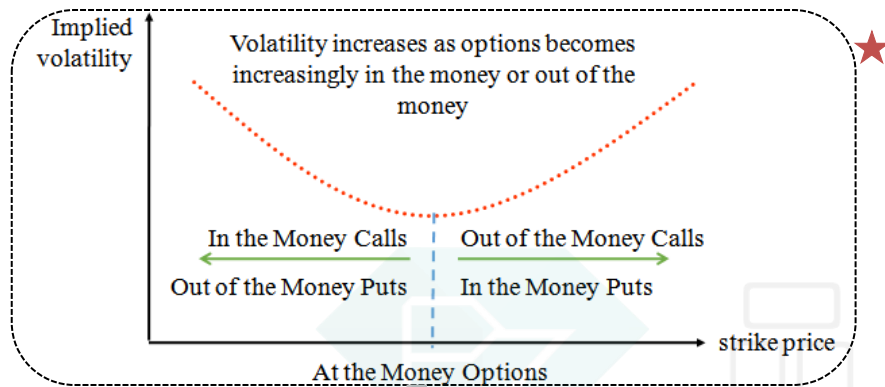
★ The implied volatility of a European call option is always the same as the implied volatility of a European put option (when the two have the same strike price and maturity date)



Alternative ways of characterizing the volatility smile

- strike price(K) $\rightarrow K/S_0 \rightarrow K/F_0$ (F_0 : the forward price of the asset)
- strike price(K) \rightarrow delta of the option

Volatility Smile For Foreign Currency Options

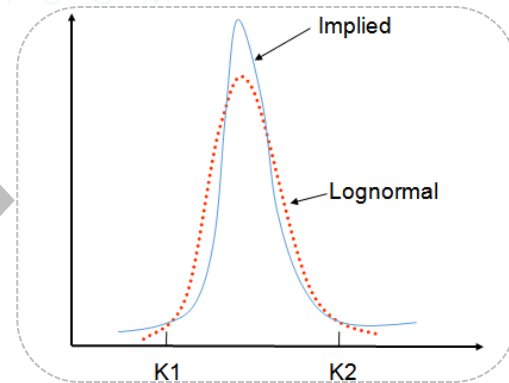


Conditions for an asset price to have a lognormal distribution

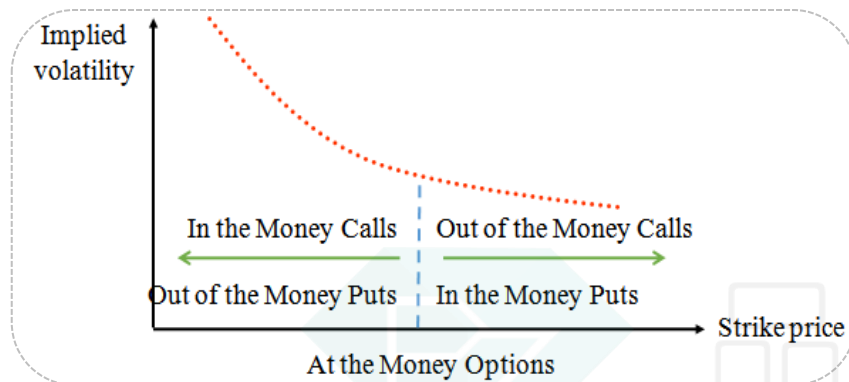
- The volatility of the asset is constant
- The price of the asset changes smoothly with no jumps

Exchange rate is not lognormally distributed

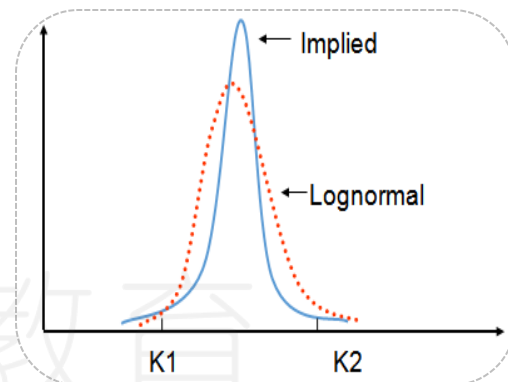
- In practice, neither of these conditions is satisfied for an exchange rate.
- The volatility of an exchange rate is far from constant, and exchange rates frequently exhibit jumps



Volatility Smiles (Skew) For Equity Options



★左高右低



Reasons For The Smile In Equity Options

- **Leverage**: firm's equity value decreases → leverage increases → increases the riskness/volatility of the underlying assets
- **Crashophobia**: 1987 stock market crash → higher premiums for put prices when the strike prices lower

The implied distribution has a heavier left tail and a less heavy right tail than the lognormal distribution.

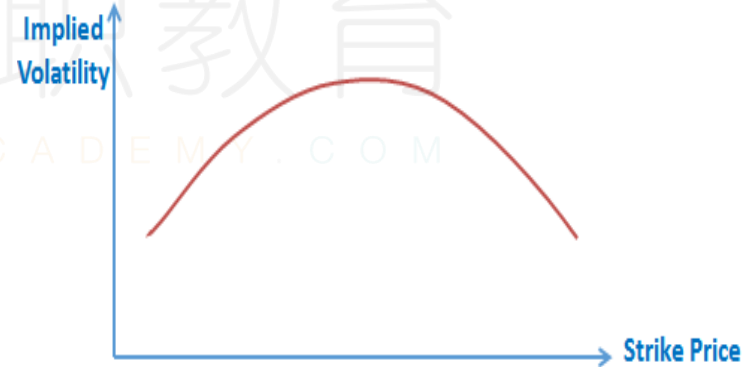
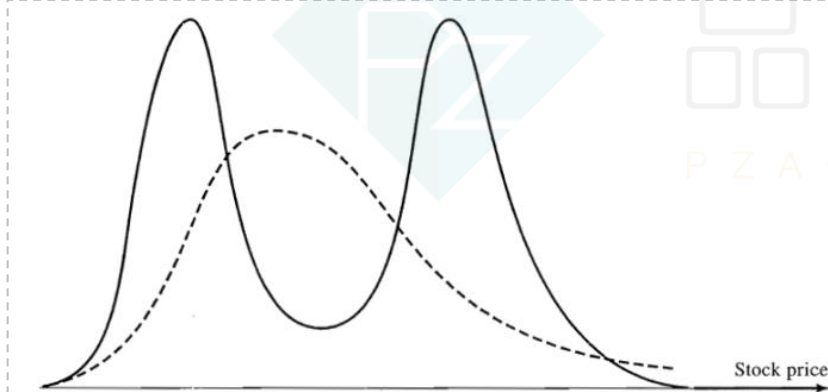
- Traders allow the implied volatility to depend on time to maturity as well as strike price.
- Volatility surfaces combine volatility smiles with the time to maturity and K/S_0 .

The Impact Of Large Asset Price Jumps

Implied volatility is affected by price jumps and the probabilities assumed for either a large up or down movement



- *At-the-money options* tend to have a higher implied volatility than either out-of-the-money or in-the-money options
- Away-from-the-money options exhibit a lower implied volatility than at-the-money options



*Thank
You!*

