

# Financial Risk Management

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Market VaR I

# Modeling Market VaR

- Parametric Models
  - Delta-Gamma
  - Monte Carlo Simulation
- Non-Parametric Simulation
  - Historical VaR
  - Semi-parametric: time and volatility weighting
- Simulations: Allocation and Confidence Intervals
- We look at 1-day VaR, unless stated otherwise

# **Model-Building / Parametric Approach**

# Model-based or Parametric VaR

1. Select a set of market variables or factors that underlie the prices and values of the portfolio
  - E.g. stock indices, interest rates, principal components
2. Assume returns of factors follow certain stochastic processes, i.e. changes in their value in the next day have certain probability distributions
  - E.g. daily stock returns are Normal
3. Estimate parameters for the underlying processes
  - E.g. use GARCH(1,1) to estimate exchange rate volatility
4. Figure out the distribution of daily changes of the portfolio based on the distribution of underlying factors
  - Closed-form or by simulation
5. Find the appropriate percentile of the distribution

# Linear-Normal Model Assumptions

- Daily change in the value of a portfolio is linearly related to the daily returns of market variables or factors
- Returns on factors are normally distributed, with mean zero, and a covariance matrix
  - Each factor return,  $i$ , has variance  $\sigma_i$
  - Every 2 factor returns,  $i$  and  $j$ , have covariance  $\text{cov}_{ij}$
- Under these assumptions, returns on the portfolio are also Normal with mean zero.
- To find VaR, we need only find the portfolio Variance

# Linear Model / Delta Method

- Define returns on market variables:  $\Delta x_i = \frac{\Delta S_i}{S_i}$
- And deltas of the portfolio with respect to asset  $i$ :

$$\delta_i = \frac{\partial P}{\partial S_i}$$

- Then changes in portfolio value are approximated by:

$$\Delta P = \sum_i S_i \delta_i \Delta x_i$$

# Variance of Portfolio Value

$$\Delta P = \sum_{i=1}^n S_i \delta_i \Delta x_i$$

$$\sigma_P^2 = \sum_{i=1}^n \sum_{j=1}^n \rho_{ij} S_i \delta_i S_j \delta_j \sigma_i \sigma_j$$

$$\sigma_P^2 = \sum_{i=1}^n (S_i \delta_i)^2 \sigma_i^2 + 2 \sum_{i < j} \rho_{ij} S_i \delta_i S_j \delta_j \sigma_i \sigma_j$$

$$\sigma_P^2 = \sum_{i=1}^n \sum_{j=1}^n \text{cov}_{ij} S_i \delta_i S_j \delta_j$$

How should we apply this method to:

1. Portfolio of Options
2. Portfolio of Bonds



# Delta Method – Example w/options

1. Consider an investment in options on Microsoft and AT&T. Suppose the stock prices are \$120 and \$30 respectively and the deltas of the portfolio with respect to the two stock prices are 1 and 20 respectively.
- Approximate the change in portfolio value as function of  $\Delta x_1$  and  $\Delta x_2$ , the returns on the two stocks:

$$\Delta P = 120 \cdot 1 \cdot \Delta x_1 + 30 \cdot 20 \cdot \Delta x_2$$

# Delta Method Example - Cont

2. Assume daily return volatility for Microsoft is 2% and that of AT&T is 1%, correlation between the two is 0.3, what is the 5-day 95% VaR?

- The variance of the portfolio is:

$$\sigma_p^2 = (120 \times 0.02)^2 + (600 \times 0.01)^2 + 2 \times 120 \times 0.02 \times 600 \times 0.01 \times 0.3 = 50.40$$

- The five-day 95% VaR is

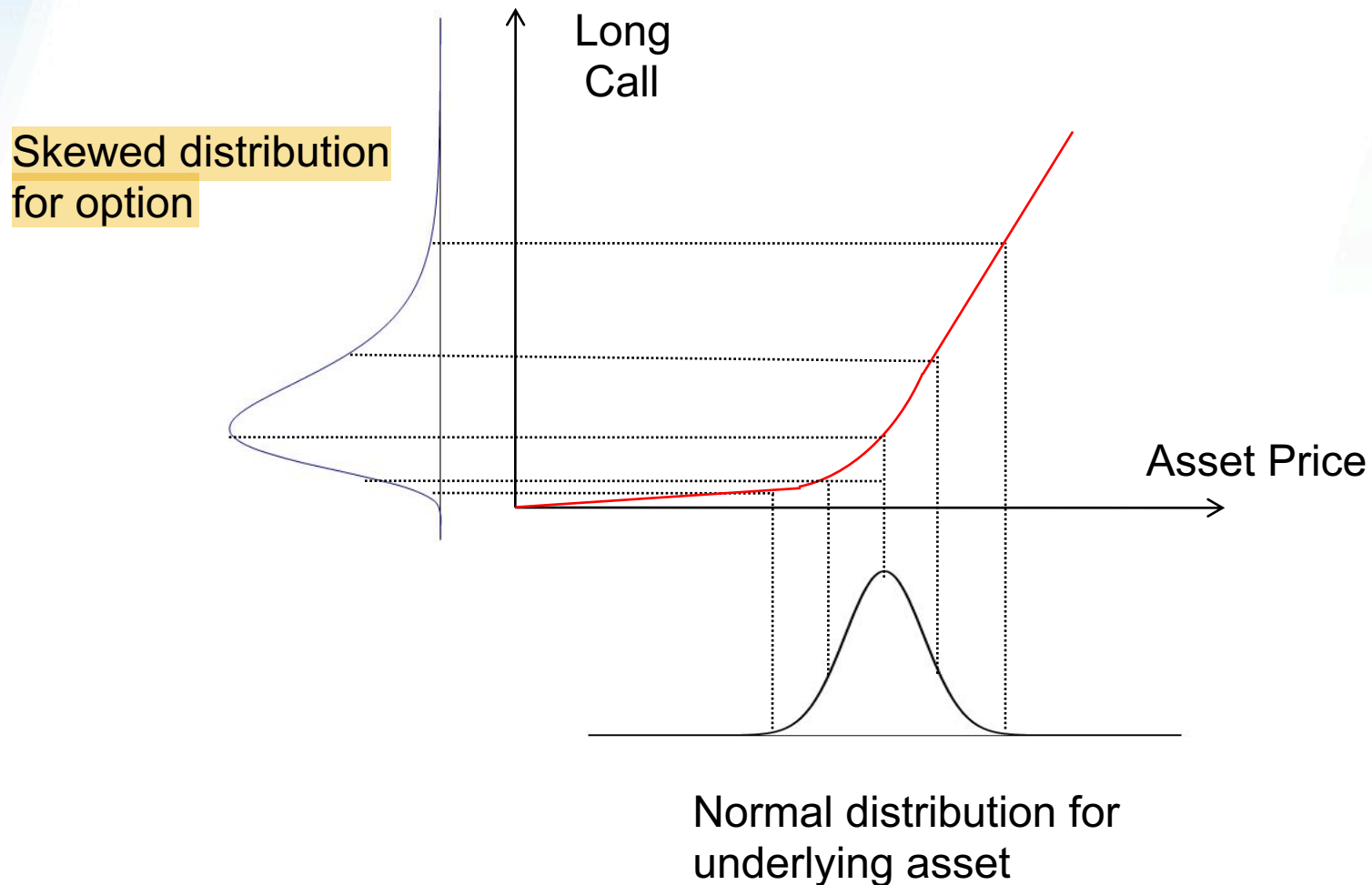
$$VaR = \Phi^{-1}(0.95) \times \sqrt{T} \times \sigma_p = 1.65 \times \sqrt{5} \times \sqrt{50.4} = 26,193$$

# Delta – Gamma

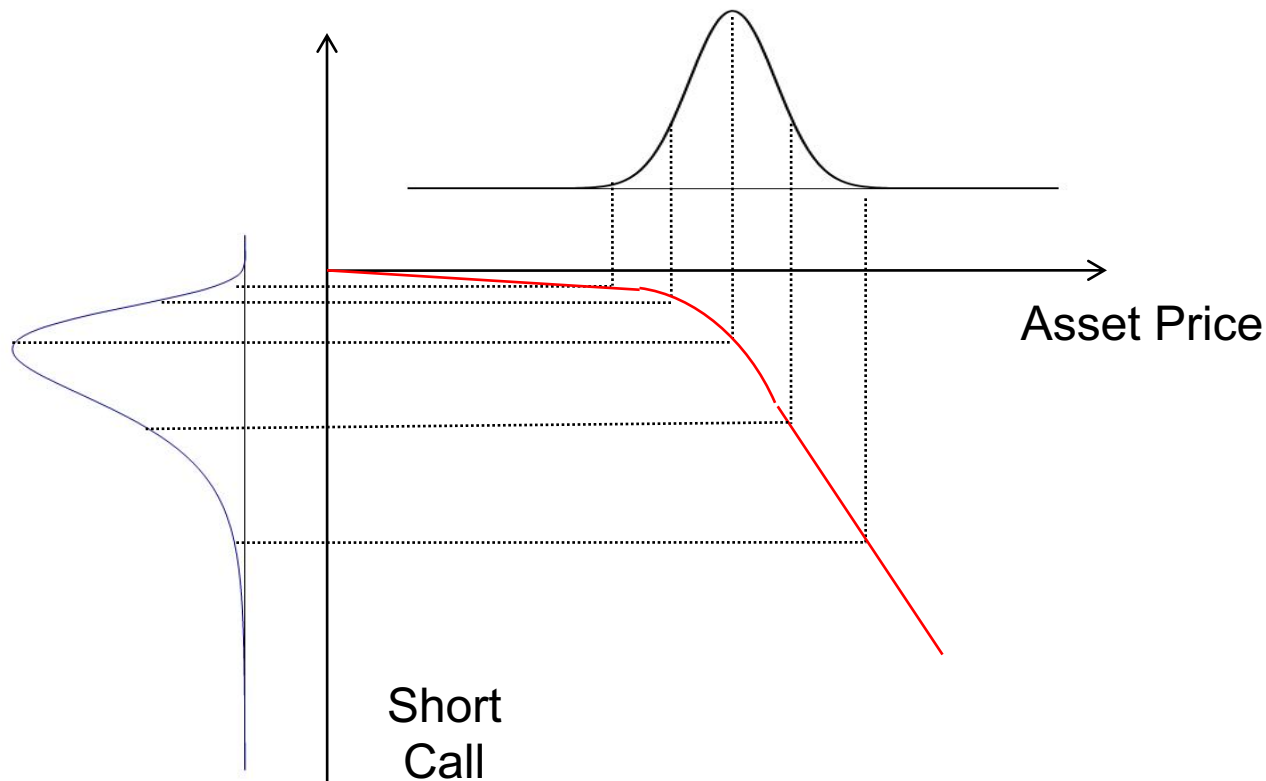
- The linear model will not be accurate because option prices are linear only for small changes in the underlying.
- We can improve our estimation by using gamma as well:

$$\Delta P \approx \delta \cdot \Delta S + \frac{1}{2} \gamma \cdot (\Delta S)^2$$

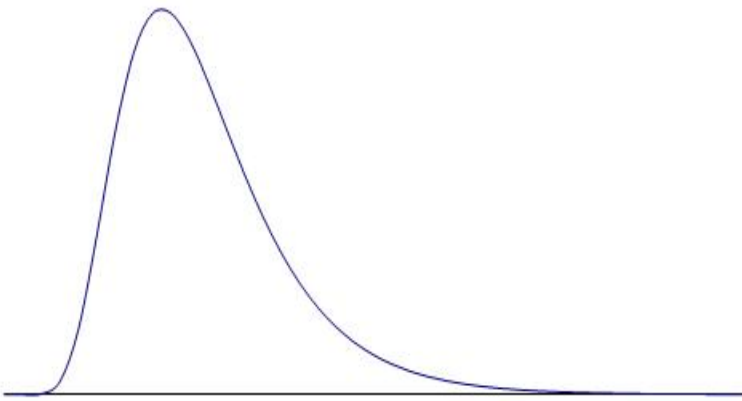
# Translation of Asset Price Changes to Price Changes for Long Call



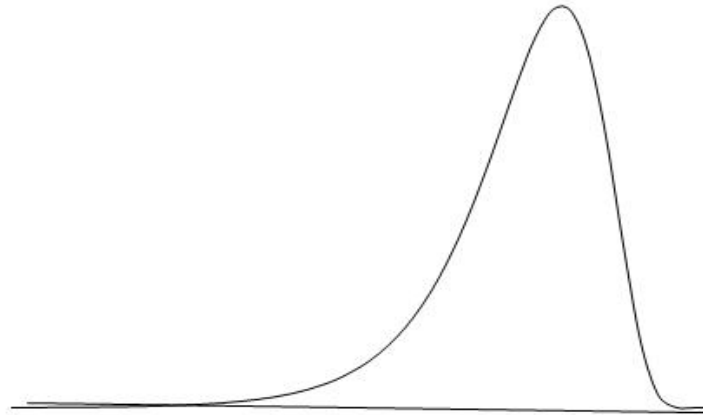
# Translation of Asset Price Change to Price Change for Short Call



# Impact of Gamma



Positive Gamma



Negative Gamma

# Quadratic / Delta-Gamma Model

- For a portfolio dependent on a single asset price it is approximately true that

$$\Delta P = \delta \Delta S + \frac{1}{2} \gamma (\Delta S)^2$$

- so that

$$\Delta P = S \delta \Delta x + \frac{1}{2} S^2 \gamma (\Delta x)^2$$

- Recall  $\Delta x \sim N(0, \sigma^2)$ , hence:

$$\mu_P = E(\Delta P) = 0.5 S^2 \gamma \sigma^2$$

$$E(\Delta P^2) = S^2 \delta^2 \sigma^2 + 0.75 S^4 \gamma^2 \sigma^4$$

$$\text{Var}(\Delta P) = E(\Delta P^2) - E(\Delta P)^2 = \delta^2 S^2 \sigma^2 + \frac{1}{2} \gamma^2 (S^2 \sigma^2)^2 = \delta^2 \text{Var}(\Delta S) + \frac{1}{2} \gamma^2 [\text{Var}(\Delta S)]^2$$

$$E(\Delta P^3) = 4.5 S^4 \delta^2 \gamma \sigma^4 + 1.875 S^6 \gamma^3 \sigma^6$$

# Quadratic Model

- When there are a small number of underlying market variable moments can be calculated analytically from the delta/gamma approximation
- The Cornish – Fisher expansion can then be used to convert moments to quantiles



# Quadratic Model – Estimating Quantiles

- Use Moments to find skewness:

$$\xi_P = \frac{1}{\sigma_P^3} E[(\Delta P - \mu_P)^3] = \frac{E[(\Delta P)^3] - 3E[(\Delta P)^2]\mu_P + 2\mu_P^3}{\sigma_P^3}$$

- Cornish – Fisher: The  $q$  percentile is  $\mu_p + w_q \sigma_p$

$$\text{where } w_q = z_q + \frac{1}{6}(z_q^2 - 1)\xi_p$$

$z_q$  is the relevant quantile of the standard normal.

For example, if we're looking for  $\text{VaR}_{95\%}$  then it will be  $N^{-1}(0.05) = -1.645$ .

# Delta-Gamma Example

- The current value of a stock index is \$1,500, and its daily volatility is 2%
- A portfolio of options on the index has  $\text{delta}=0.5$ ,  $\text{gamma}=0$
- Use the delta-gamma method to estimate the mean, variance and skewness of dollar returns of the portfolio.
- What is 1-day VaR-95%?

# Delta-Gamma Example

- Mean=0
- Variance= $\delta^2 S^2 \sigma^2 = (0.5)^2 * 1500^2 * (0.02)^2 = 225$
- Skewness=0
- The 5%-ile is:  $0 - 1.645 \cdot \sqrt{225} = -25$
- VaR-95% =  $|-25| = 25$

# Effect of Gamma on Portfolio and VaR

What if Portfolio Gamma = 0.07? Gamma = -0.07?

Gamma	0.07	0	-0.07
$E(\Delta P)$	31.5	0	-31.5
$\text{var}(\Delta P)$	2,210	225	2,210
skewness	2.817	0.000	-2.817
$w_q$	-0.844	-1.645	-2.446
5%-ile	-8	-25	-146

# Modeling Bonds in Linear Model

- Duration Approach: Linear relation between  $\Delta P$  and  $\Delta y$  (allows parallel shifts only)
- Zero Coupon Bonds: Underlying variables are zero-coupon bond returns with many different maturities
- Principal Components Approach: 2 or 3 independent shifts with their own volatilities, capture most of the variance in term-structure moves

# Duration Approach

- Recall  $\frac{\Delta P}{P} \sim -Duration \cdot \Delta y$
- Therefore:  $\Delta P \approx -P \cdot Duration \cdot \Delta y = -Dollar\ Duration \cdot \Delta y$
- Assume yield follows:  $\Delta y \sim N(\mu_y, \sigma_y^2)$
- Then the price will follow:  $\Delta P \sim N(\mu_P, \sigma_P^2)$

$$\text{s.t.} \quad \mu_P = -DD \cdot \mu_y \quad \sigma_P = DD \cdot \sigma_y$$

$$VaR_{95\%} = -(\mu_P - 1.645 \times \sigma_P) = DD \cdot (\mu_y + 1.645 \times \sigma_y)$$

# Duration Approach

- Suppose that the volatility of daily changes in interest rates is 0.1% with mean=0
- Our Portfolio is worth \$820 and has duration of 5
- Using the normal-linear approach find the 1-day  $\text{VaR}_{95\%}$ :
  - The dollar duration is  $820 \times 5 = 4,100$ 
    - Portfolio value will drop by \$41 for 1% rise in yield.
  - $\text{VaR}_{95\%} = 4100 \times [0 + 1.645 \times 0.001] = \$6.74$

# Duration Approach Caveats

- The duration approximation is for small changes in yield, VaR might involve large changes, where the approximation fails
  - Include convexity and use delta-gamma.
- The portfolio might be affected by non-parallel shifts in the yield curve.



# Zero-coupon Bond Returns as Underlying Variables

- We can choose as market variables zero-coupon bond price changes with standard maturities (for example: 1m, 3m, 6m, 1yr, 2yr, 5yr, 7yr, 10yr, 30yr)
- We need to estimate the covariance matrix of all these bond price returns.
- We need to map the portfolio to each of the maturities.
- Suppose we have  $n$  maturities:

$$\sigma_P^2 = \sum_{i=1}^n \sum_{j=1}^n \text{cov}_{ij} P_i P_j$$

# Bond Portfolio Example

- Consider a portfolio invested \$37,397 in 3mm, \$331,382 in 6mm and \$678,074 in 1-yr bonds
- Rates, vols and correlations for bond prices:

	3-Month	6-Month	1-Year
Zero rate (% with ann. comp.)	5.50	6.00	7.00
Bond price vol (% per day)	0.06	0.10	0.20
Correlation between daily returns			
	3-Month Bond	6-Month Bond	1-Year Bond
3-month bond	1.0	0.9	0.6
6-month bond	0.9	1.0	0.7
1-year bond	0.6	0.7	1.0

## Example – cont.

Portfolio Variance =

$$\begin{aligned} & 37,397^2 * (0.06\%)^2 + 331,382^2 * (0.10\%)^2 + \\ & 678,074^2 * (0.20\%)^2 + 2 * 37,397 * 331,382 * (0.06\%) * \\ & (0.10\%) * 0.9 + 2 * 331,382 * 678,074 * (0.10\%) * (0.20\%) * 0.7 \\ & + 2 * 37,397 * 678,074 * (0.06\%) * (0.20\%) * 0.6 = 2,628,518 = 1,621.3^2 \end{aligned}$$

10-day VaR-99% =

$$1621.3 * 2.33 * \text{sqrt}(10) = \$11,946$$

# Zero-Coupon Bond Return Disadvantages

- Requires many underlying variables
- We need to map bond cash flows that arrive at times different than our underlying bonds.

# Using PCA to Calculate VaR

- We can use 2 or 3 PCAs as underlying factors.
- It requires:
  - Portfolio sensitivities to those factors
  - Volatilities of the factors
- We estimate less deltas and don't need covariance matrix as PCs are orthogonal

**Table 8.8** Standard Deviation of Factor Scores

PC1	PC2	PC3	PC4	PC5	PC6	PC7	PC8
17.55	4.77	2.08	1.29	0.91	0.73	0.56	0.53

# PCA Example

Suppose a portfolio has the following sensitivities to 1-basis-point rate moves, in \$ millions:

3-Year Rate	4-Year Rate	5-Year Rate	7-Year Rate	10-Year Rate
+10	+4	-8	-7	+2

What are the portfolio sensitivities to PC1 and PC2?

$$\text{PC1: } 10 \times 0.372 + 4 \times 0.392 - 8 \times 0.404 - 7 \times 0.394 + 2 \times 0.376 = -0.05$$

$$\text{PC2: } 10 \times (-0.267) + 4 \times (-0.110) - 8 \times 0.019 - 7 \times 0.194 + 2 \times 0.371 = -3.87$$

# PCA Example – Cont.

- We get:

$$\Delta P = -0.05f_1 - 3.87f_2$$

where  $f_1$  is the first factor and  $f_2$  is the second factor

- If the SD of the factor scores are 17.55 and 4.77 the SD of  $\Delta P$  is

$$\sqrt{0.05^2 \times 17.55^2 + 3.87^2 \times 4.77^2} = 18.48$$



# Thanks