

# DOWNSIDE RISK – IMPLICATIONS FOR FINANCIAL MANAGEMENT

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# RISK AND RETURN

- THE TRADE-OFF BETWEEN RISK AND RETURN IS THE CENTRAL PARADIGM OF FINANCE.
- HOW MUCH RISK AM I TAKING?
- HOW SHOULD I RESPOND TO RISKS THAT VARY OVER TIME?
- HOW SHOULD I RESPOND TO RISKS OF VARIOUS MATURITIES?



# DOWNSIDE RISK

- THE RISK OF A PORTFOLIO IS THAT ITS VALUE WILL DECLINE, NOT THAT IT WILL INCREASE HENCE DOWNSIDE RISK IS NATURAL.
- MANY THEORIES AND MODELS ASSUME SYMMETRY: c.f. MARKOWITZ, TOBIN, SHARPE AND VOLATILITY BASED RISK MANAGEMENT SYSTEMS.
- DO WE MISS ANYTHING IMPORTANT?

# MEASURING DOWNSIDE RISK

- Many measures have been proposed. Let  $r$  be the one period continuously compounded return with distribution  $f(r)$  and mean zero. Let  $x$  be a threshold.

- Skewness =  $E(r^3) / E(r^2)^{3/2}$
- Probability of loss =  $P(r < x)$ ,
- Expected loss =  $E(r | r < x)$
- $x$  is the  $\alpha$  Value at risk if  $P(r < -x) = \alpha$

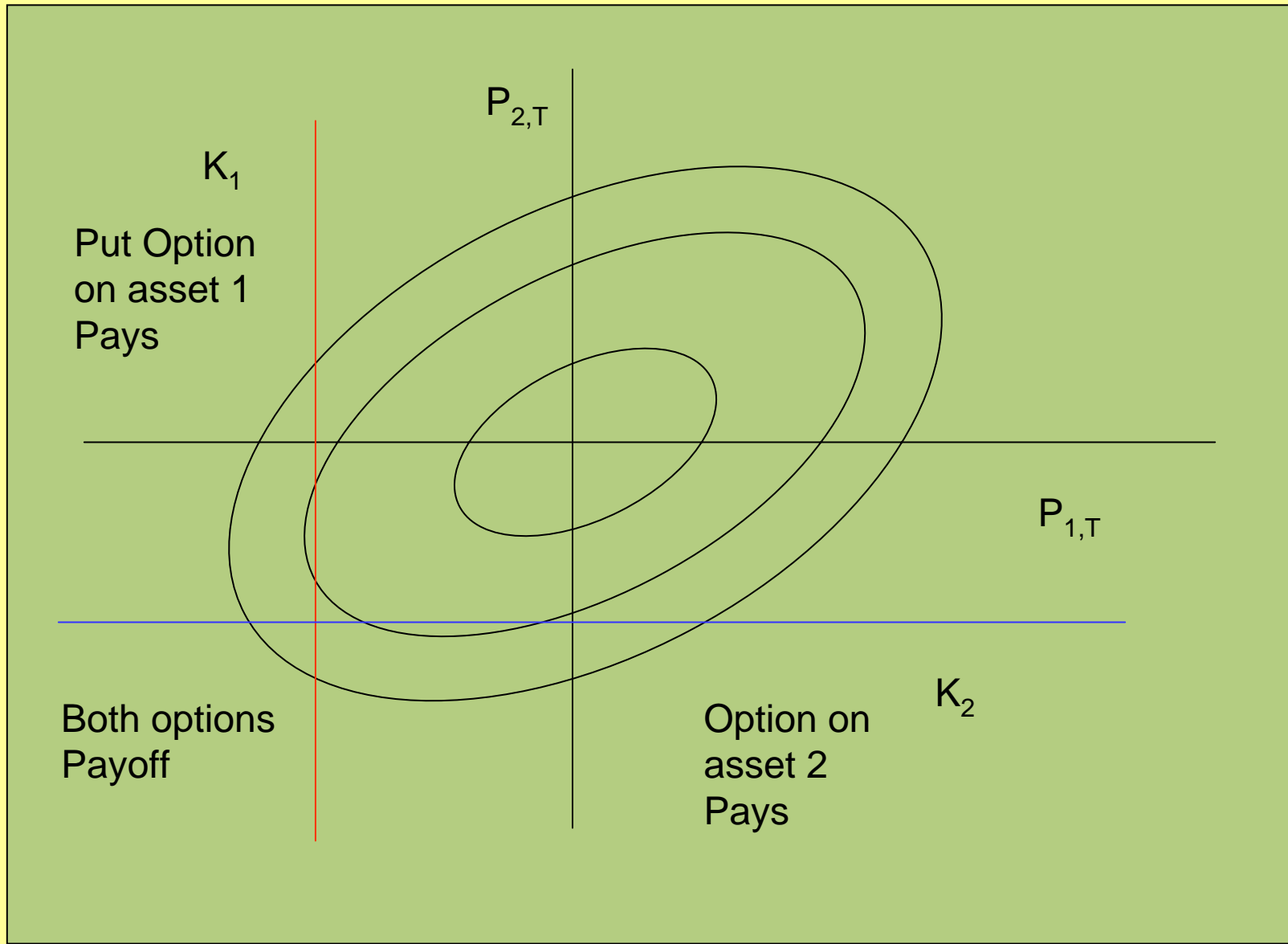


# MULTIVARIATE DOWNSIDE RISK

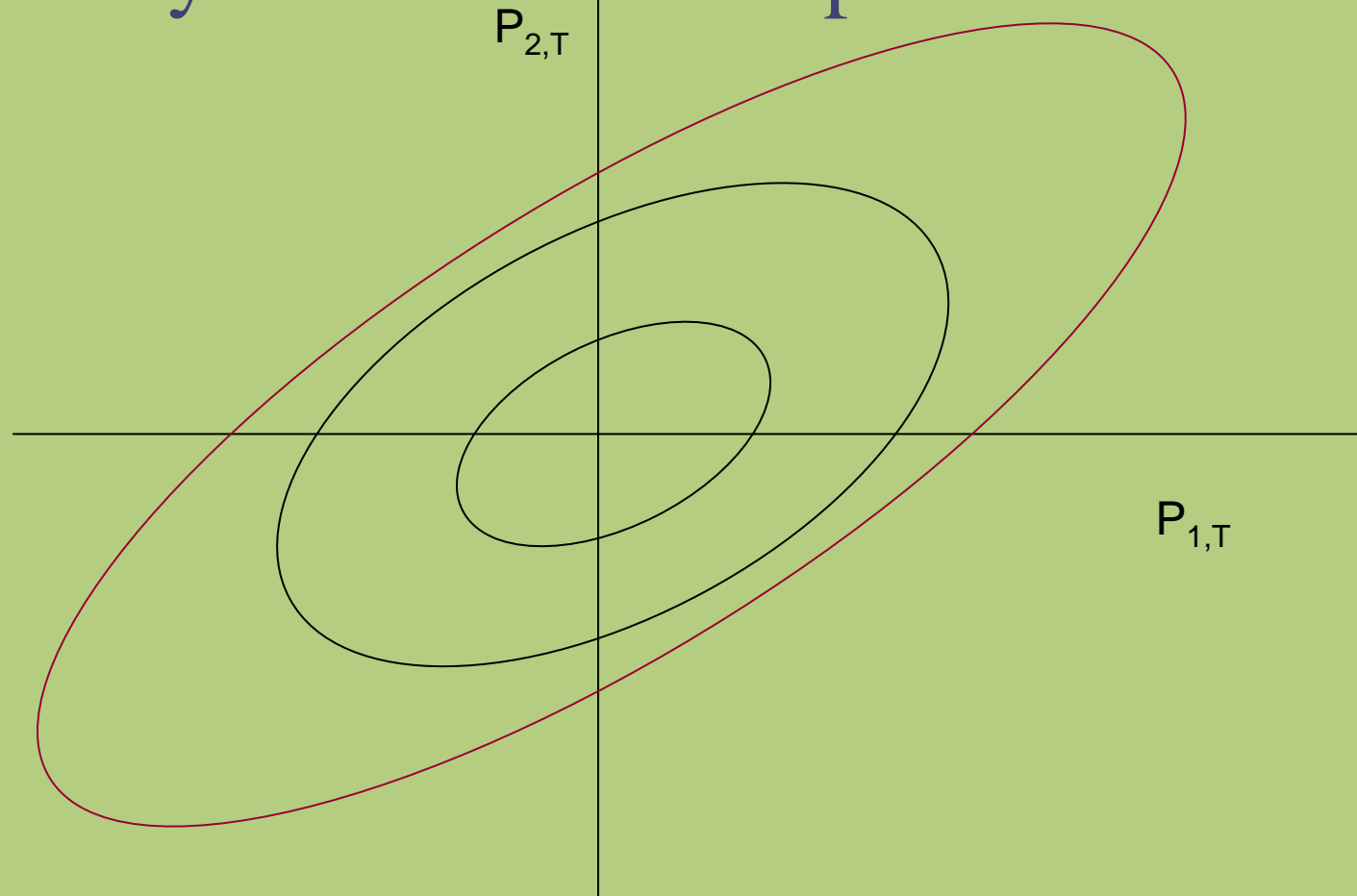
- WHAT IS THE LIKELIHOOD THAT A COLLECTION OF ASSETS WILL ALL DECLINE?
- THIS DEPENDS PARTLY ON CORRELATIONS
- FOR EXTREME MOVES, OTHER MEASURES ARE IMPORTANT TOO.

$$W_1P_1+W_2P_2=-K$$

Probability  
that the  
portfolio  
loses more  
than  $K$

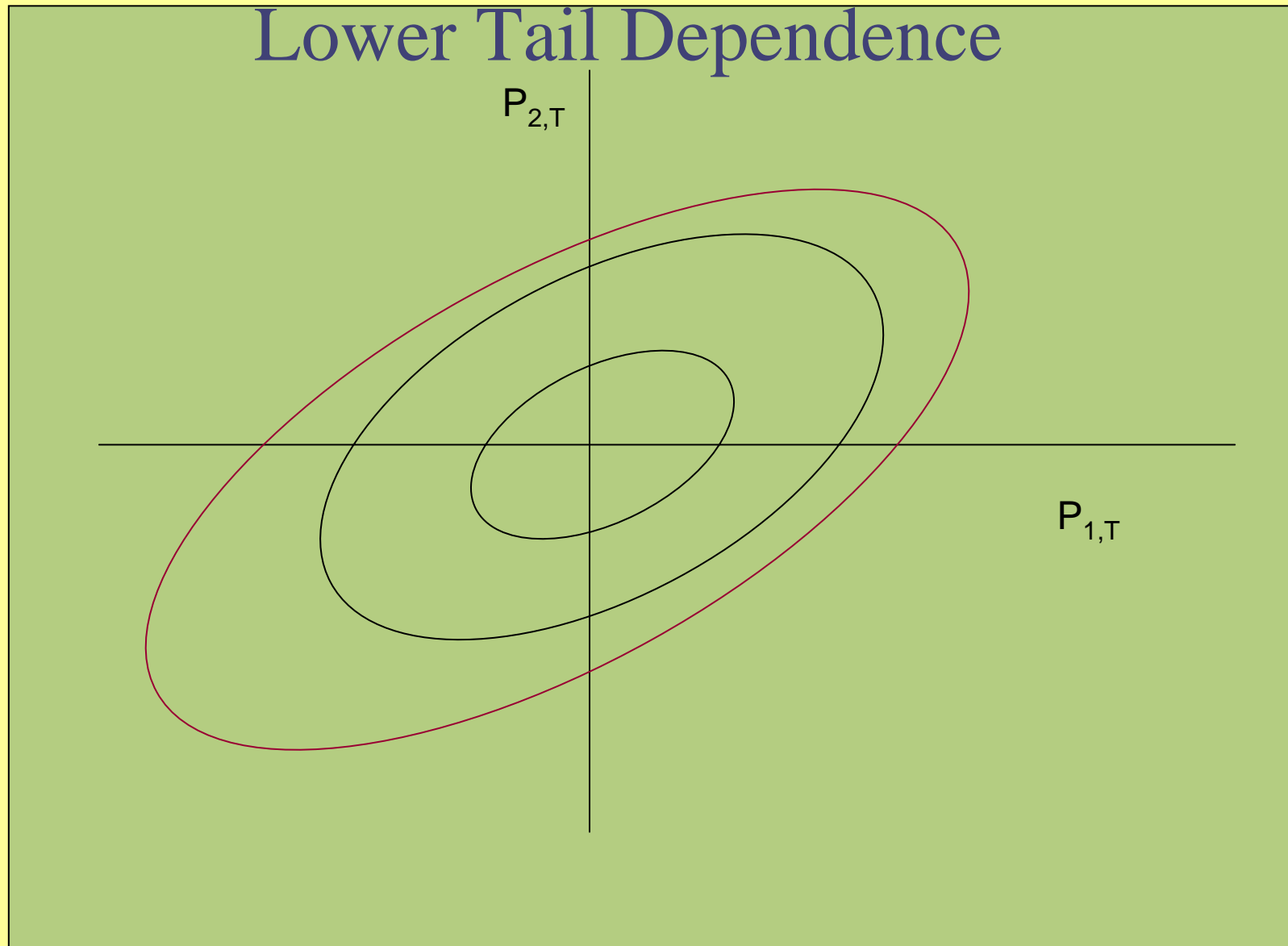


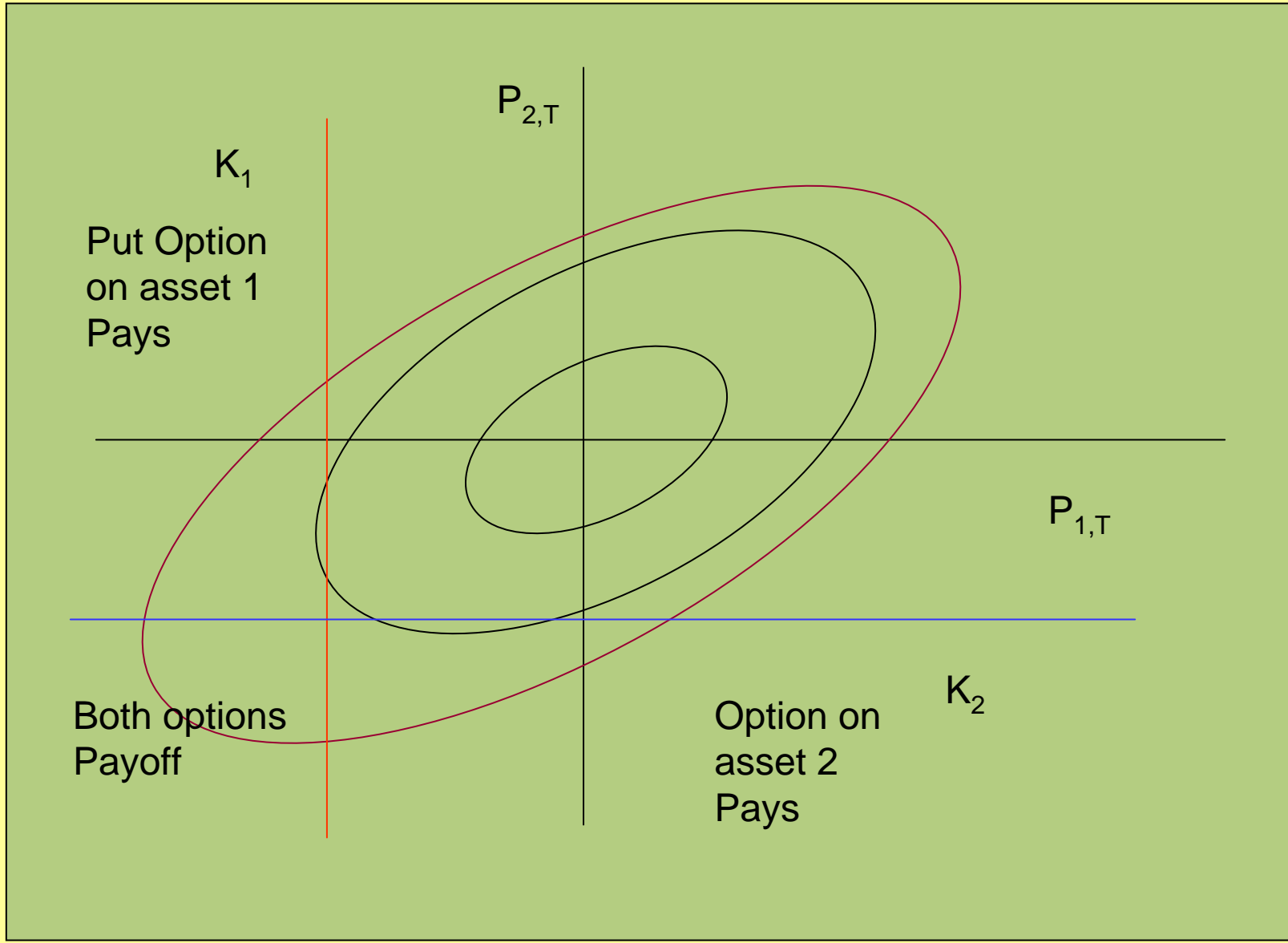
# Symmetric Tail Dependence





# Lower Tail Dependence







# CONTAGION

- WHERE ARE MY CORRELATIONS WHEN I NEED THEM?
- WHEN COUNTRIES DECLINE TOGETHER MORE THAN CAN BE EXPECTED FROM THE NORMAL CORRELATION PATTERN, IT IS CALLED CONTAGION.
- CORRELATIONS AND VOLATILITIES APPEAR TO MOVE TOGETHER.



# CREDIT DERIVATIVES

- IT IS WELL DOCUMENTED THAT THE MULTIVARIATE NORMAL DENSITY UNDERPRICES JOINT EXTREME EVENTS SUCH AS DEFAULTS.
- INDUSTRY HAS ADOPTED A T-COPULA TO PRICE CREDIT BASKETS and CDO's.
- TAIL DEPENDENCE IS ESSENTIAL IN THESE MODELS.

# THE PURPOSE OF MY TALK TODAY

TIME SERIES ANALYSIS OF  
DOWNSIDE RISK



# PURPOSE OF MY TALK TODAY

- TO SHOW HOW DOWNSIDE RISK CAN BE MODELED AS A TIME SERIES PROCESS
- USING SIMPLY TIME AGGREGATION OF STANDARD TIME SERIES MODELS
- *CONSEQUENTLY*
- DOWNSIDE RISK CAN BE PREDICTED
- DYNAMIC HEDGING AND DYNAMIC PORTFOLIO STRATEGIES CAN BE IMPLEMENTED.

# AN ECONOMETRIC FRAMEWORK

- MODEL THE ONE PERIOD RETURN AND CALCULATE THE MULTI-PERIOD CONDITIONAL DISTRIBUTION

- RETURN FROM  $t$  UNTIL  $t + T$  IS:

$$R_t^T = \sum_{j=t+1}^{T+t} r_j$$

- THE DISTRIBUTION CONDITIONAL ON TODAY'S INFORMATION IS:

$$R_t^T | \mathcal{F}_t \sim f_t^T (R_t^T)$$



## ALL MEASURES CAN BE DERIVED FROM THE ONE PERIOD DENSITY

- EVALUATE ANY MEASURE BY  
REPEATEDLY SIMULATING FROM THE  
ONE PERIOD CONDITIONAL  
DISTRIBUTION:

$$f_t(r_{t+1})$$

- METHOD:
  - Draw  $r_{t+1}$
  - Update density and draw observation t+2
  - Continue until T returns are computed.
  - Compute measure of downside risk



# A MODEL

- $r_t = \sqrt{h_t} \varepsilon_t, \varepsilon_t \sim i.i.d.$

$$E_{t-1}(r_t) = 0, h_t = V_{t-1}(r_t)$$

- ASYMMETRY FOLLOWS FROM ASYMMETRY IN EPSILON
- HOWEVER FOR MULTI-PERIOD RETURNS, THERE IS ANOTHER SOURCE – ASYMMETRIC VOLATILITY.



# The ARCH Model

- The ARCH model of Engle(1982) is a family of specifications for the conditional variance.
- The  $q^{\text{th}}$  order ARCH or ARCH( $q$ ) model is

$$h_t = \omega + \sum_{j=1}^q \alpha_j r_{t-j}^2$$

- Notice that it is a simple generalization of both constant variance and rolling variance estimates called “historical volatilities”.

# GARCH

- The Generalized ARCH model of Bollerslev(1986) is an ARMA version of this model. GARCH(1,1) is a weighted average of three volatility forecasts:

$$h_t = \omega + \alpha r_{t-1}^2 + \beta h_{t-1}$$

# Asymmetric Volatility

- Often negative shocks have a bigger effect on volatility than positive shocks.
- Nelson(1987) introduced the EGARCH model to incorporate this effect.
- I will use a Threshold GARCH or TARARCH

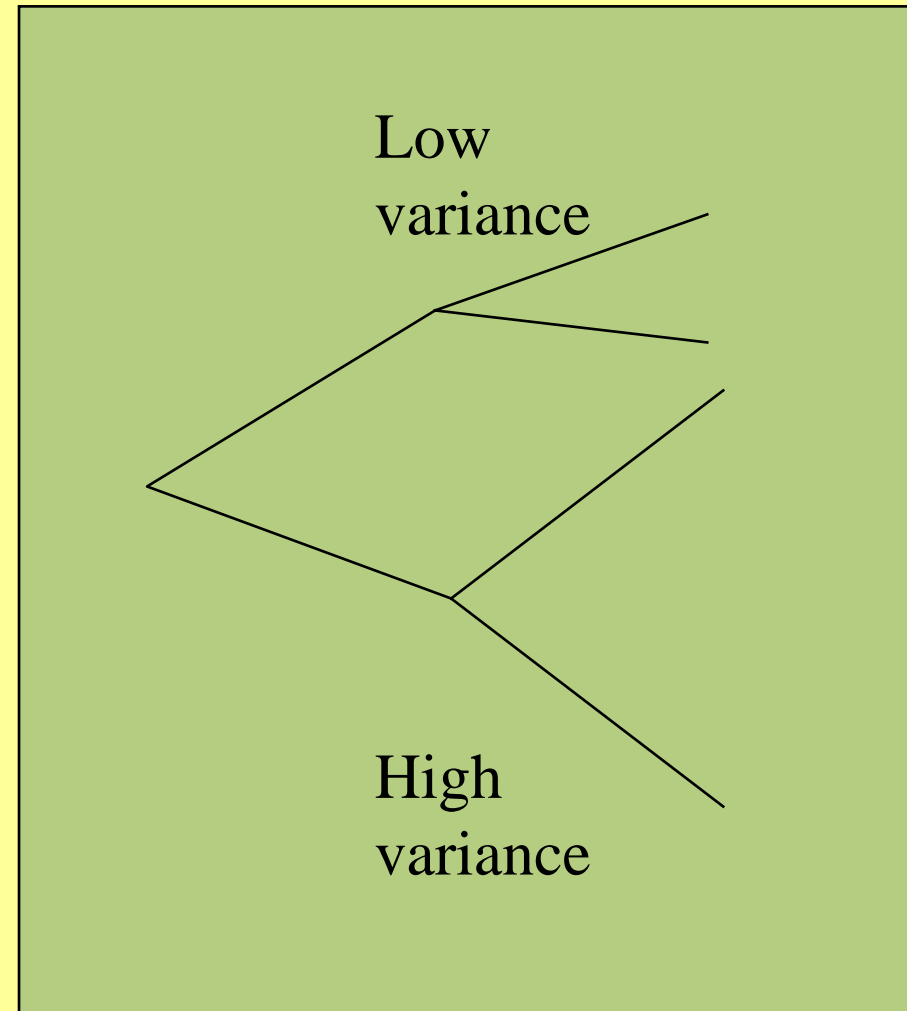
$$h_t = \omega + \alpha r_{t-1}^2 + \gamma r_{t-1}^2 I_{(r_{t-1} < 0)} + \beta h_{t-1}$$

# NEW ARCH MODELS

- GJR-GARCH
- TARCH
- STARCH
- AARCH
- NARCH
- MARCH
- SWARCH
- SNPARCH
- APARCH
- TAYLOR-SCHWERT
- FIGARCH
- FIEGARCH
- Component
- Asymmetric Component
- SQGARCH
- CESGARCH
- Student t
- GED
- SPARCH
- Autoregressive Conditional Density
- Autoregressive Conditional Skewness

# TWO PERIOD RETURNS

- Two period return is the sum of two one period continuously compounded returns
- Look at binomial tree version
- Asymmetric Volatility gives negative skewness



# ANALYTICALLY: TARARCH WITH SYMMETRIC INNOVATIONS

$$\begin{aligned} E\left(r_t + r_{t+1}\right)^3 &= E\left(r_t^3 + 3r_t^2 r_{t+1} + 3r_t r_{t+1}^2 + r_{t+1}^3\right) \\ &= 0 + 0 + 3E\left(r_t h_{t+1}\right) + 0 \\ &= 3E\left[r_t\left(\omega + \alpha r_t^2 + \gamma r_t^2 I_{(r_t < 0)} + \beta h_t\right)\right] \\ &= 3\gamma E\left(r_t^3 I_{(r_t < 0)}\right) < 0 \end{aligned}$$

and the conditional third moment is

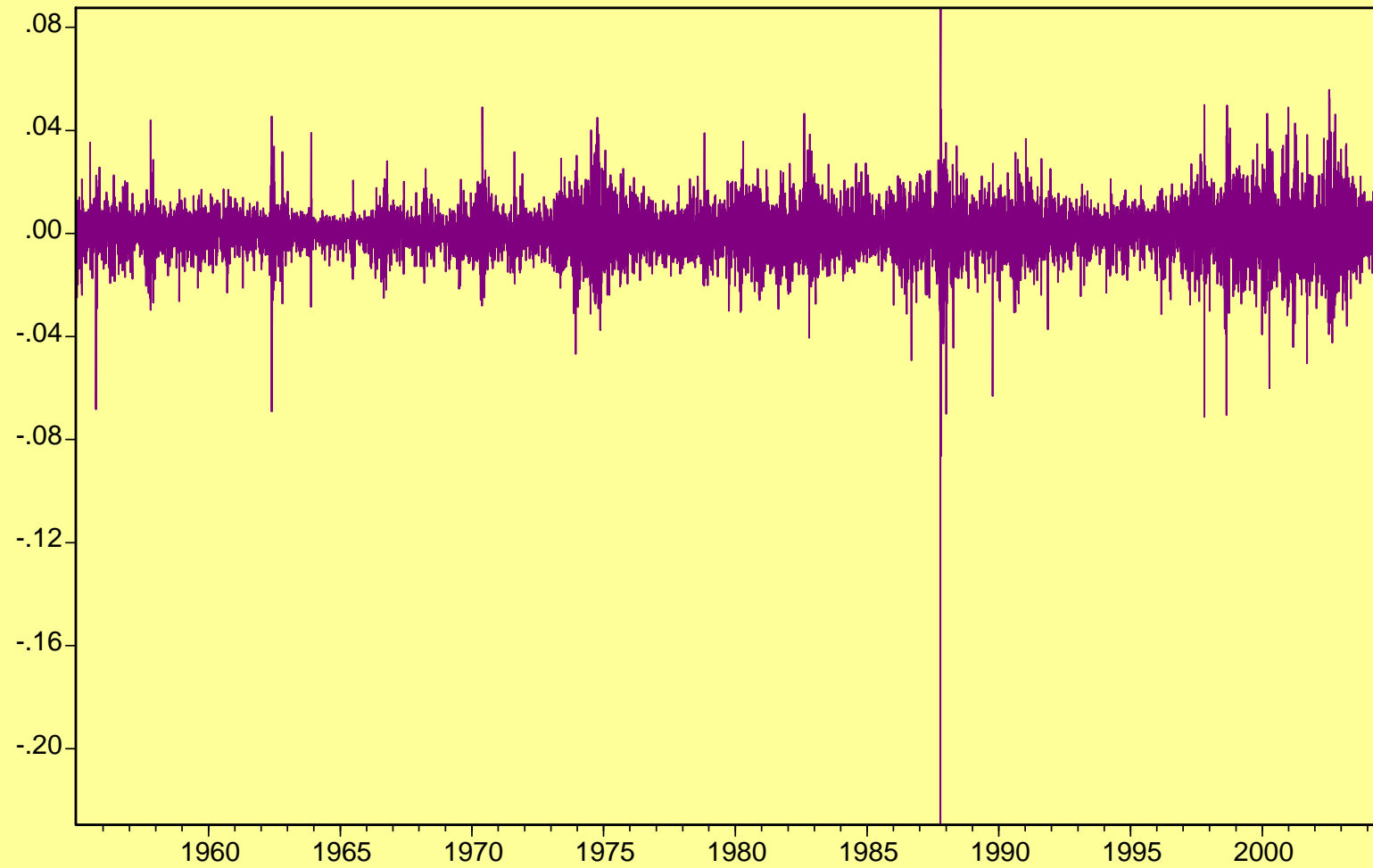
$$E_{t-1}\left(r_t + r_{t+1}\right)^3 = 3\gamma E_{t-1}\left(r_t^3 I_{(r_t < 0)}\right) < 0$$

# STYLIZED FACTS

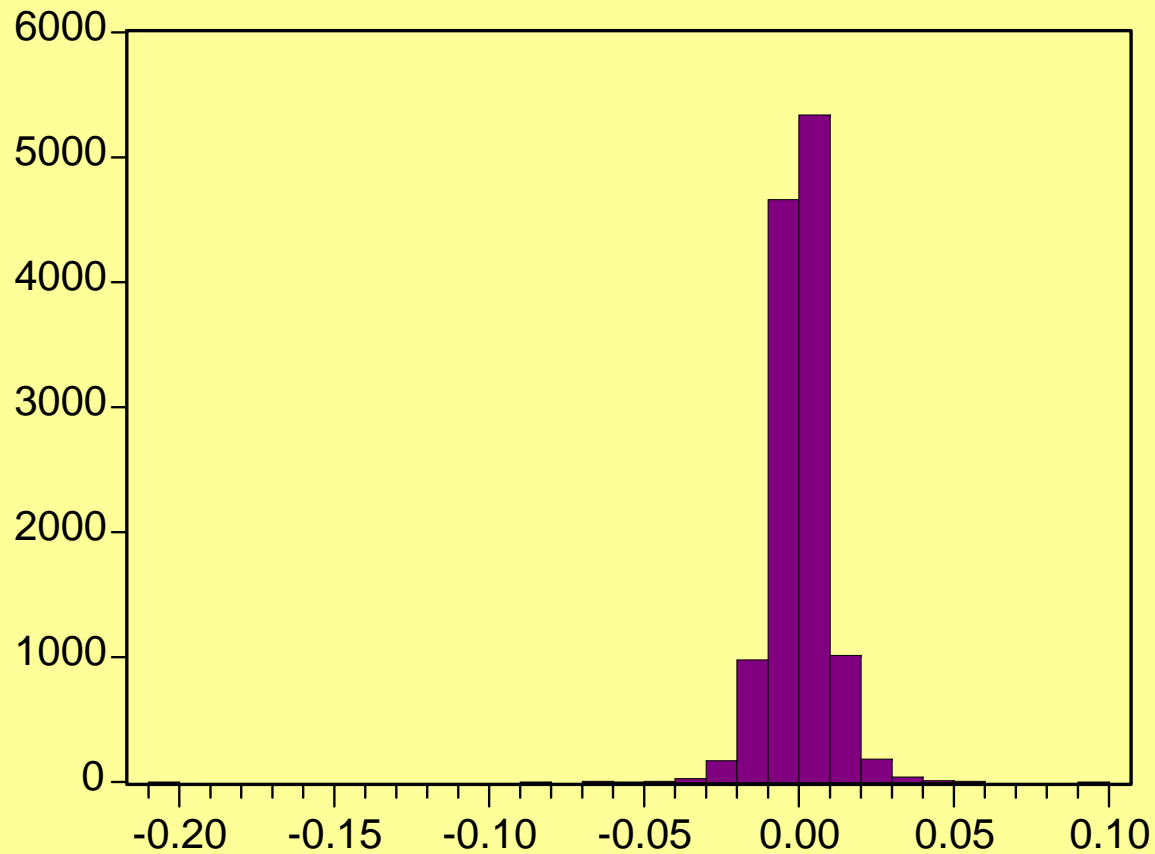




# S&P 500 DAILY RETURNS



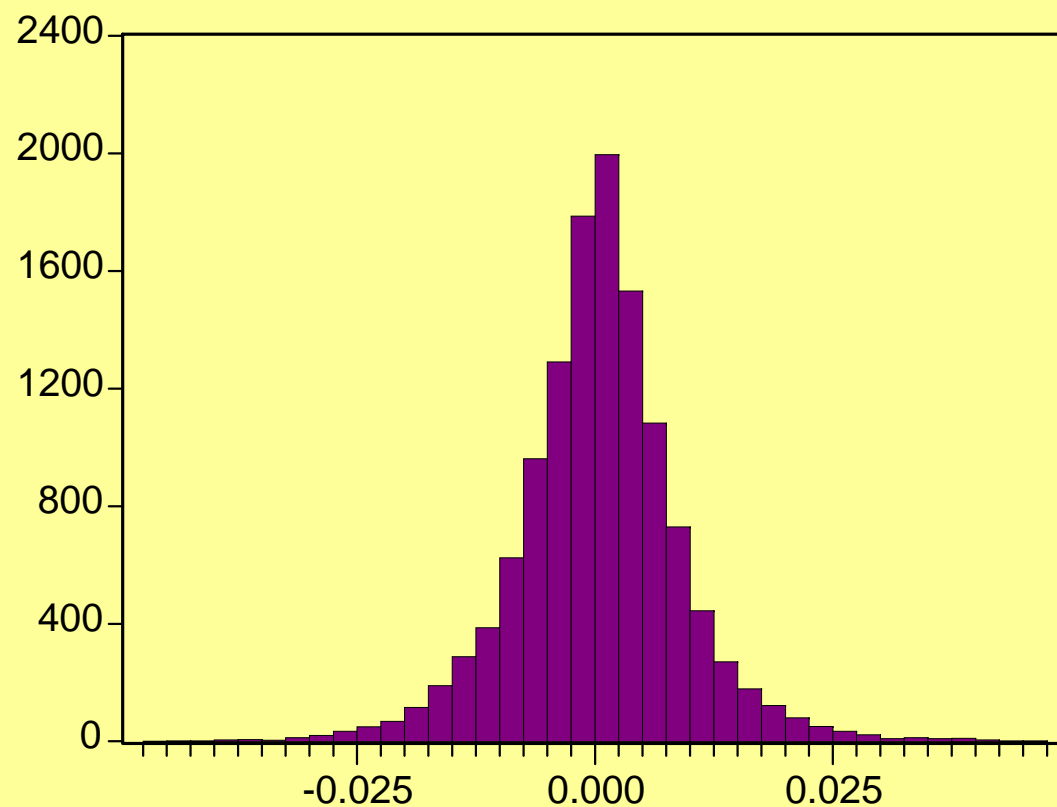
HISTOGRAM OF S&P500 DAILY RETURNS



Series: RETUSSP	
Sample 1/4/1955 TO 6/25/2004	
Observations 12455	
Mean	0.000318
Median	0.000375
Maximum	0.090994
Minimum	-0.204669
Std. Dev.	0.009179
Skewness	-0.926286
Kurtosis	28.00273
Jarque-Bera	326200.8
Probability	0.000000

# TRIMMING .001 IN EACH TAIL

(8 DAYS)

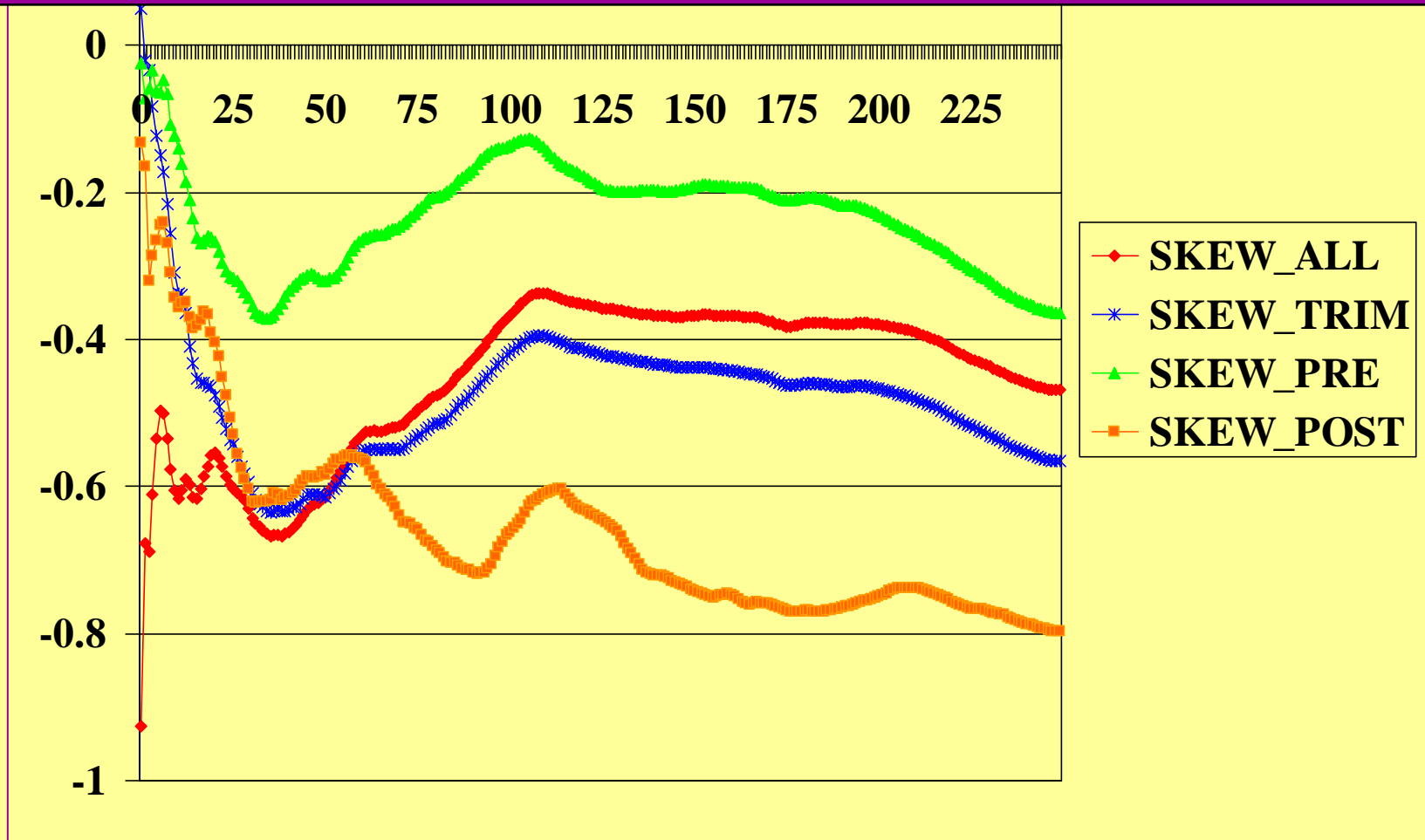


Series: RETUSSP  
Sample 1 12455 IF Y>LOWTRIM  
AND Y<HIGHTRIM  
Observations 12431

Mean	0.000338
Median	0.000375
Maximum	0.046486
Minimum	-0.045594
Std. Dev.	0.008633
Skewness	0.049617
Kurtosis	5.380668

Jarque-Bera	2940.670
Probability	0.000000

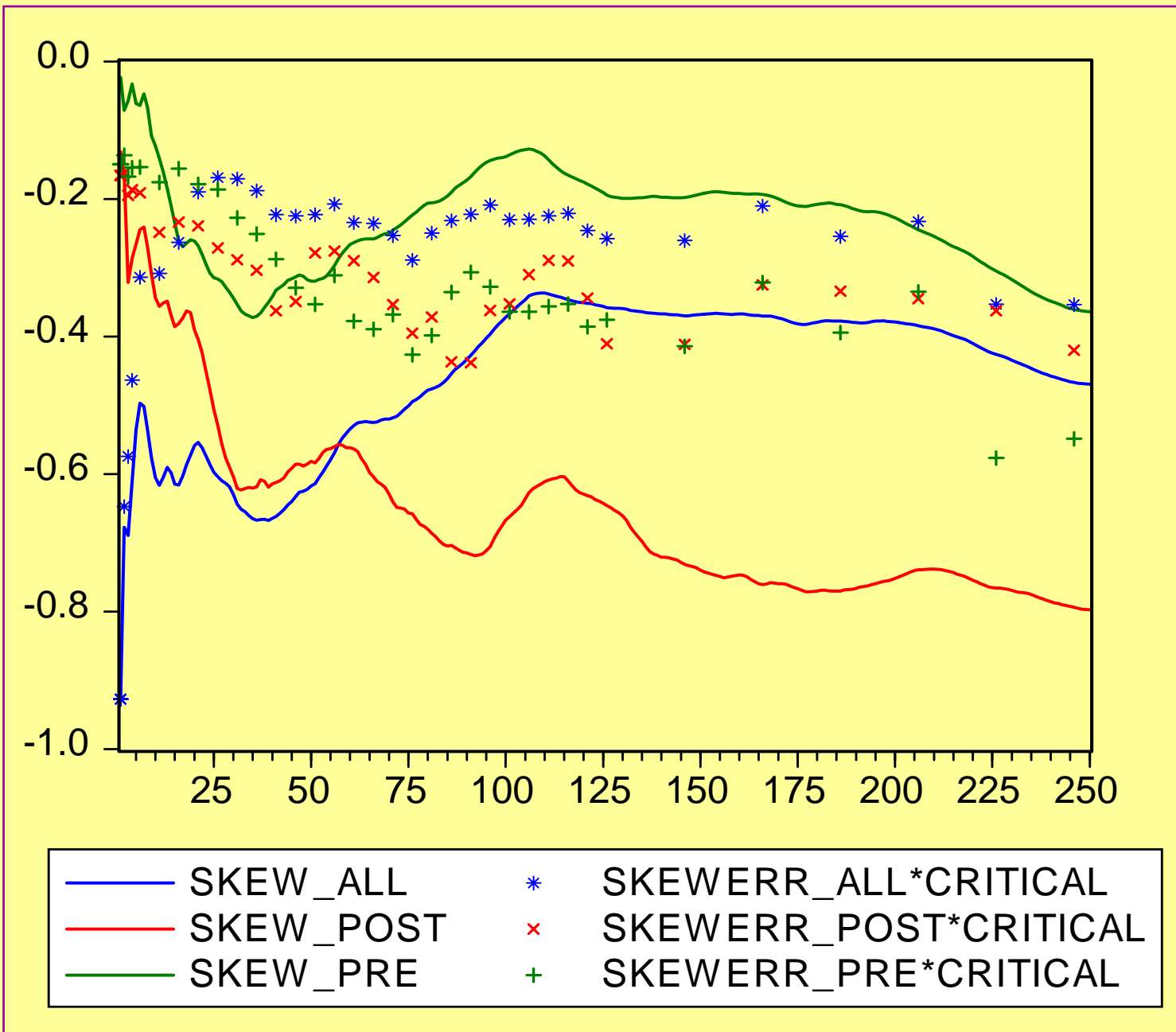
# SKEWNESS OF MULTIPERIOD RETURNS





# STANDARD ERRORS

- ARE THESE DIFFERENCES SIGNIFICANT?
- THE INFERENCE IS COMPLICATED BY THE OVERLAPPING OBSERVATIONS AND BY THE DEPENDENCE DUE TO ESTIMATING THE MEAN.
- FROM SIMPLE ROBUST TESTS, SIZE CORRECTED BY MONTE CARLO, THESE ARE SIGNIFICANT.





# EVIDENCE FROM DERIVATIVES

- THE HIGH PRICE OF OUT-OF-THE-MONEY EQUITY PUT OPTIONS IS WELL DOCUMENTED
- THIS IMPLIES SKEWNESS IN THE RISK NEUTRAL DISTRIBUTION
- MUCH OF THIS IS PROBABLY DUE TO SKEWNESS IN THE EMPIRICAL DISTRIBUTION OF RETURNS.
- DATA MATCHES EVIDENCE THAT THE OPTION SKEW IS ONLY POST 1987.

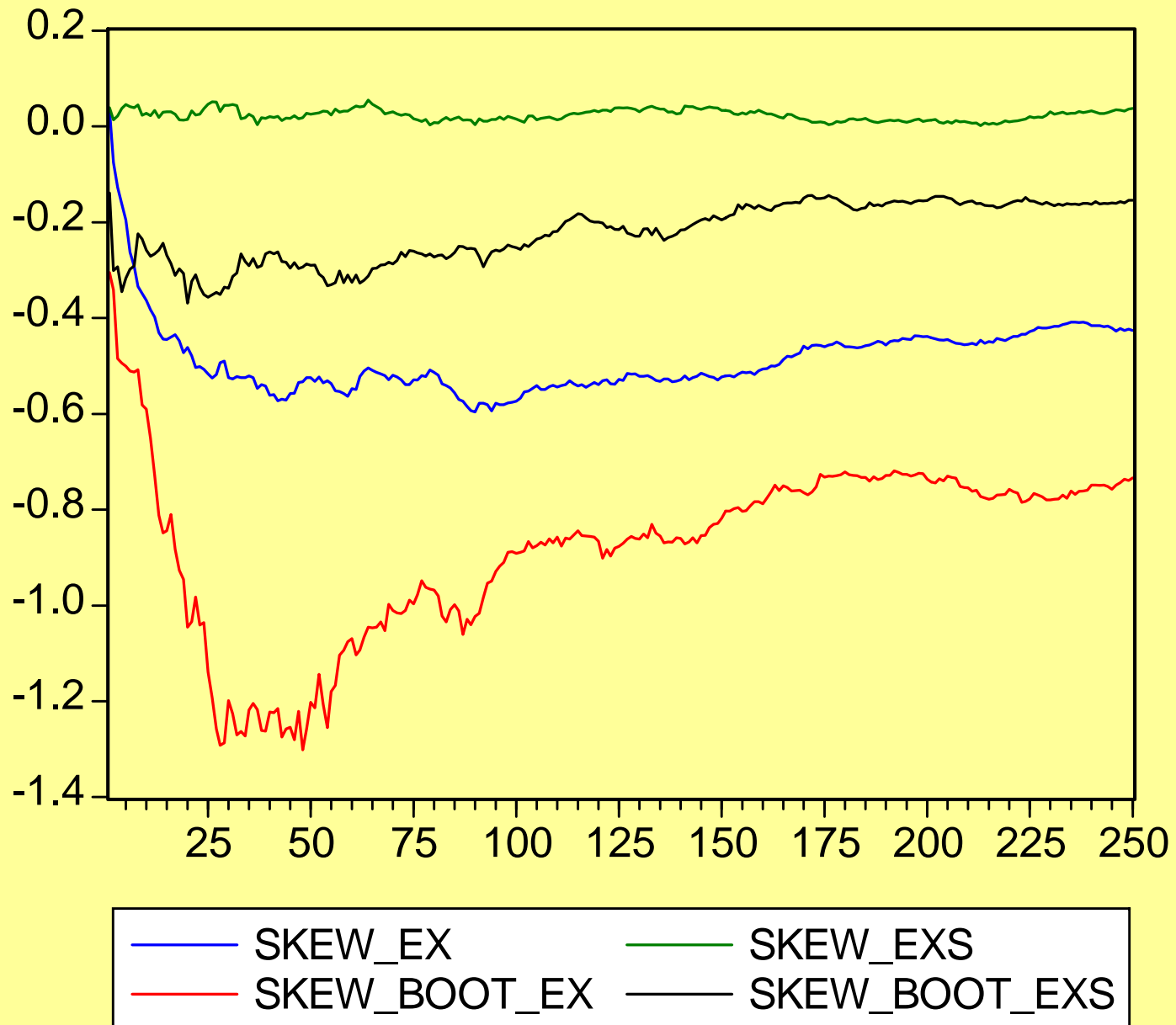


# MATCHING THE STYLIZED FACTS

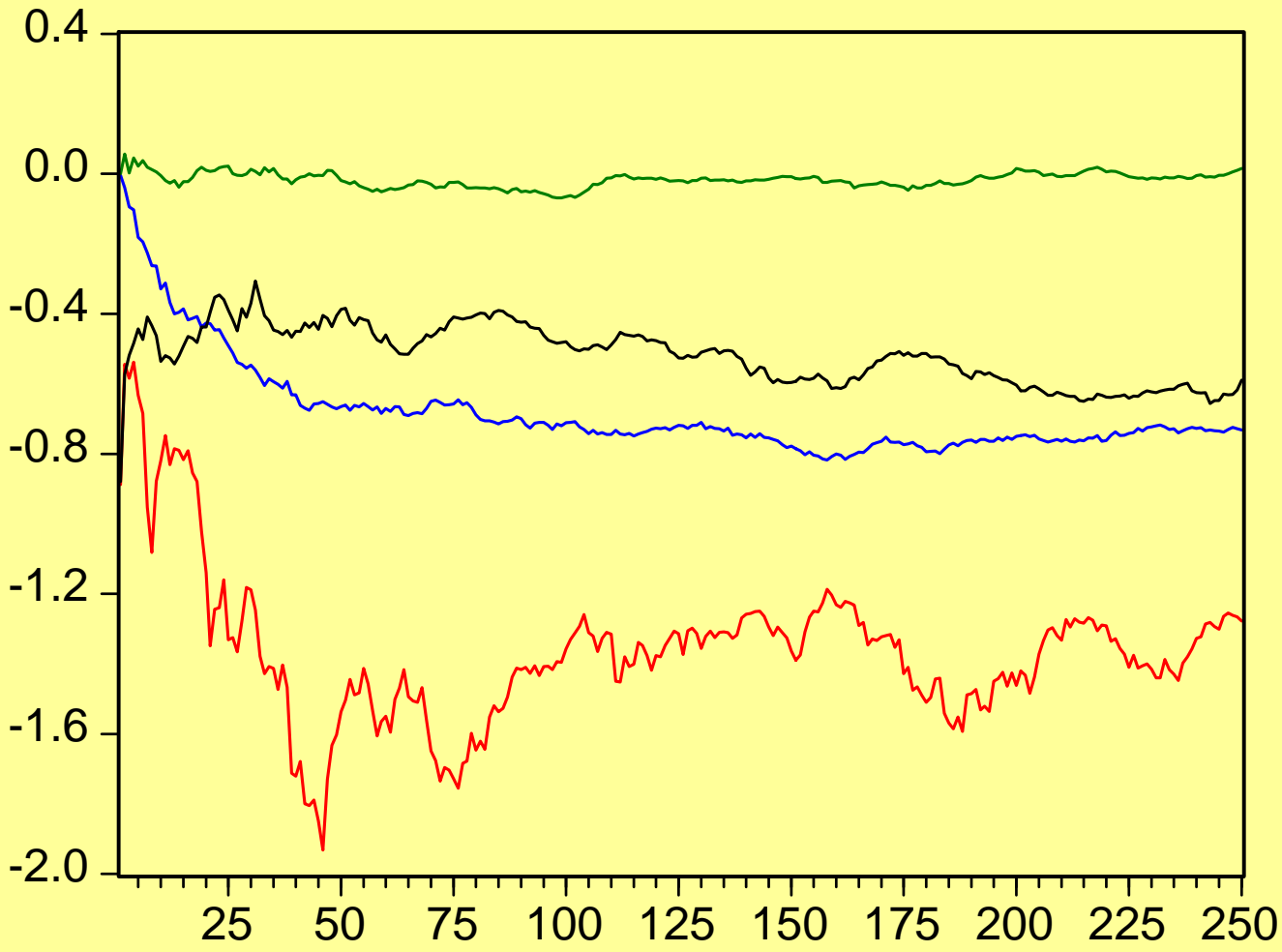
- ESTIMATE DAILY MODEL
- SIMULATE 250 CUMULATIVE RETURNS 10,000 TIMES WITH SEVERAL DATA GENERATING PROCESSES
- CALCULATE SKEWNESS AT EACH HORIZON



# SKEWS FOR SYMMETRIC AND ASYMMETRIC MODELS



# SIMULATED SKEWNESS FROM 1988



— SKEW\_TARCH\_88      — SKEW\_GARCH\_88  
— SKEW\_BOOT\_TARCH\_88      — SKEW\_BOOT\_GARCH\_88



# IMPLICATIONS

- Multi-period empirical returns are more skewed than one period returns (omitting 1987 crash)
- Asymmetric volatility is needed to explain this.
- Skewness has increased since 1987, particularly for longer horizons.
- Simulated skewness is noisy because higher moments do not exist when the persistence is so close to one. Presumably this is true for the data too.
- Many other asymmetric models could be compared on this basis.

# MULTIVARIATE MODELS





# DOWNSIDE RISK RESULTS FROM TIME AGGREGATION WITH:

## ● ASYMMETRIC CORRELATIONS

- CORRELATIONS RISE PARTICULARLY AFTER TWO ASSETS BOTH DECLINE. (Asymmetric DCC (Cappiello, Engle, Sheppard(2004))

## ● VOLATILITY SHOCKS ARE CORRELATED

- PURE VARIANCE COMMON FEATURES(Engle, Marcucci(2005))
- FACTOR MODELS (Engle Ng and Rothschild(1992))
- CREDIT RISK MODEL(Engle, Berd, Voronov(2005))

# FACTOR ARCH

- RETURNS ARE DRIVEN BY A SMALL NUMBER OF FACTOR SHOCKS,  $f_t$ .
- FACTORS DRIVE VOLATILITIES AND CORRELATIONS

$$r_t = Bf_t + u_t$$

$$V_{t-1}(r_t) = B\Omega_t B + D_t$$

$$V_{t-1}(f_t) = \Omega_t, V_{t-1}(u_t) = D_t$$

# DOWNSIDE RISK IN THE CAPM

- The return on a stock can be decomposed into systematic and idiosyncratic returns using the beta of the stock

- $$r_{i,t} = \beta_i r_{m,t} + \varepsilon_{i,t}$$

- If the market declines substantially, many stocks will decline. There will be skewness in each stock and downside risk in the portfolio.

# SKEWNESS

- Under the standard assumptions, the skewness of return  $i$  is related to the return of the market by  $s_i = s_m R^3$  where  $R^3$  is the conventional  $R^2$  raised to the  $3/2$  power.
- Notice that all stocks will then have skewness but that it will be less than for the market.



# TAIL DEPENDENCE

- The probability that two stocks will both underperform some threshold can be calculated conditional on the market return.

$$\begin{aligned} P(r_i < k \text{ and } r_j < k) &= E\left(P(r_i < k \text{ and } r_j < k | r_m)\right) \\ &= E\left(P(\varepsilon_i < k - \beta_i r_m | r_m) P(\varepsilon_j < k - \beta_j r_m | r_m)\right) \\ &= P(r_i < k) P(r_j < k) \\ &\quad + Cov\left(P(\varepsilon_i < k - \beta_i r_m | r_m), P(\varepsilon_j < k - \beta_j r_m | r_m)\right) \\ &= E\left(\Phi(k - \beta_i r_m) \Phi(k - \beta_j r_m)\right) \text{ under normality} \end{aligned}$$



# SUMMARY

- ASYMMETRIC VOLATILITY IN THE MARKET FACTOR IMPLIES
  - SKEWNESS IN MULTIPERIOD MARKET RETURNS
  - SKEWNESS IN MULTIPERIOD EQUITY RETURNS
  - LOWER TAIL DEPENDENCE IN EQUITY RETURNS

# IMPLICATIONS FOR FINANCIAL MANAGEMENT





# IMPLICATIONS FOR RISK MANAGEMENT

- MULTI-PERIOD RISKS MAY BE SUBSTANTIALLY DIFFERENT FROM ONE PERIOD RISKS.
- THE MULTI-PERIOD RISK CHANGES OVER TIME AND CAN BE FORECAST.
- BIG MARKET DECLINES ARE MORE LIKELY WHEN VOLATILITY IS HIGH



# IMPLICATIONS FOR DERIVATIVE HEDGING

- AS EACH NEW PERIOD RETURN IS OBSERVED, THE DERIVATIVE CAN BE REPRICED AND THE HEDGE UPDATED.
- GREEKS CAN BE CALCULATED FROM SIMULATION PRICING TO SIMPLIFY THE UPDATING



# IMPLICATIONS FOR PORTFOLIO SELECTION

- MEAN VARIANCE PORTFOLIO OPTIMIZATION WILL MISS THESE ASYMMETRIES.
- HIGH FREQUENCY REBALANCING WILL GIVE *EARLY WARNING* OF DOWNSIDE RISK.



# HOW TO DO THIS?

## ● SUBOPTIMAL METHOD 1

- MYOPIC ASSET ALLOCATION ON A HIGH FREQUENCY BASIS.
- AS VOLATILITIES RISE, YOU NATURALLY SHIFT OUT OF RISKY ASSETS.

## ● SUBOPTIMAL METHOD 2

- MULTI-PERIOD FORECAST OF RISK GIVES AN EX-ANTE OPTIMAL PLAN.
- OVERINVEST WHEN VOLATILITY IS LOW AND UNDERINVEST WHEN IT IS HIGH



# OPTIMAL METHOD

## ● DYNAMIC PROGRAMMING:

- WHEN VOLATILITY IS LOW, UNDERINVEST, RECOGNIZING THAT THIS PLAN MAY CHANGE WHEN THE SUBSEQUENT VOLATILITY IS OBSERVED
- SEE COLACITO AND ENGLE(2004)





# EXPECTED RETURNS

- EACH OF THESE METHODS REQUIRES EXPECTED RETURNS.
- THE LISTED IMPLICATIONS ARE BASED ON THE ASSUMPTION THAT EXPECTED RETURNS ARE UNCHANGED.
- IS THIS REASONABLE?



## BUT IF EVERYBODY DID THIS?

- IF ALL AGENTS FOLLOW THIS PATTERN THEN EXPECTED RETURNS WOULD NECESSARILY ADJUST. RETURNS WOULD INSTANTANEOUSLY MOVE ENOUGH TO RESTORE EQUILIBRIUM. CAMPBELL AND HENTSCHEL(1992)
- IN A REPRESENTATIVE AGENT WORLD, THERE WOULD NO LONGER BE A MOTIVE FOR ADJUSTING TO CHANGES IN RISK.
- CHANGES IN RISK WOULD LEAD TO UNAVOIDABLE CAPITAL GAINS OR LOSSES.
- DERIVATIVE REPLICATION STRATEGIES WOULD CONTINUE TO BE USEFUL.



# HOWEVER

- THERE IS NO REASON TO BELIEVE DOWNSIDE RISK WOULD DISAPPEAR OR COLLAPSE TO AN INSTANT IN TIME.
- WITH HETEROGENEITY, THERE WOULD STILL BE REASONS TO REBALANCE.
- FROM A MICROSTRUCTURE POINT OF VIEW IT IS DIFFICULT TO IMAGINE HOW THE PRICES COULD INSTANTANEOUSLY ADJUST TO VOLATILITY NEWS.
- EXPECTED RETURNS WOULD BE EXCEEDINGLY DIFFICULT TO ESTIMATE AT THIS HIGH FREQUENCY
- MAYBE WE ARE ALREADY AT THIS POINT SO THAT DOWNSIDE RISK IS FULLY AND INSTANTLY PRICED.



# CONCLUSIONS

- ASYMMETRIC VOLATILITY AND CORRELATION MODELS ARE POWERFUL TOOLS FOR ANALYZING DOWNSIDE RISK
- ONE PERIOD MODELS HAVE BIG IMPLICATIONS ABOUT LONG HORIZON OF RETURNS
- THE UPDATING OF VOLATILITY AND RISK MEASURES HAS A NATURAL APPLICATION TO DERIVATIVE HEDGING AND POSSIBLY PORTFOLIO REBALANCING.