

Stock index futures

Markets and applications



Stock index derivatives

- Purpose:
 - Describe stock index derivatives markets.
 - Apply index derivatives for managing return/risk.
- Outline:
 - History of stock index products
 - Stock index construction
 - Index arbitrage
 - Hedging stock market risk



Stock index futures

- History of listings in US
 - Value Line futures
 - Kansas City Board of Trade in February 1982
 - S&P 500 futures
 - Chicago Mercantile Exchange in April 1982
 - Major Market Index futures
 - Chicago Board of Trade in July 1984
 - DJIA futures
 - Chicago Board of Trade in October 1997



Stock index futures

- History of listings in other countries
 - All Ordinaries
 - SFE in February 1983
 - TSE 300
 - TSE in January 1984
 - FT-SE 100
 - LIFFE in May 1984
 - BOVESPA
 - BOVESPA in February 1986



Stock index futures

- Major derivatives exchanges often have futures contracts on multiple indexes.
 - “First-mover” usually gathers lion’s share of trading volume.
 - One index is preeminent.
 - US: S&P 500
 - Europe: Euro Stoxx 50



Exchange products

- Stock index futures
 - Dominant US exchange is CME Group
 - <https://www.cmegroup.com/trading/products/>

Exchange products

Product name	Index	Volume	Open interest	Turnover
<u>E-mini S&P 500 Futures</u>	US	1,466,350	2,762,774	53.1%
<u>E-mini Russell 2000 Index Futures</u>	US	116,286	490,965	23.7%
<u>S&P 500 Total Return Index Futures</u>	US	3,000	272,701	1.1%
<u>E-mini Nasdaq-100 Futures</u>	US	545,487	206,243	264.5%
<u>E-mini Dow (\$5) Futures</u>	US	238,563	104,903	227.4%
<u>S&P 500 Annual Dividend Index Futures</u>	US	3,786	92,798	4.1%
<u>E-mini S&P MidCap 400 Futures</u>	US	11,871	78,286	15.2%
<u>Micro E-mini S&P 500 Index Futures</u>	US	291,768	56,399	517.3%
<u>Nikkei/Yen Futures</u>	International	49,610	54,924	90.3%
<u>Dow Jones Real Estate Futures</u>	US	757	40,410	1.9%
<u>Micro E-mini Nasdaq-100 Index Futures</u>	US	256,361	38,119	672.5%
<u>S&P 500 Futures</u>	US	1,640	37,628	4.4%
<u>E-mini Financial Select Sector Futures</u>	Select Sector	1,267	31,965	4.0%

Highest open interest.

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Pit-traded contract is 5 times size.

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Variety of different stock indexes

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Major US stock indexes, international indexes, sector indexes

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Low values are buy-and-hold futures.
High values are day-trading futures.



Exchange products

- Supporting file: Index return statistics.xlsx
 - Downloaded daily data from Datastream for period 20091231 through 20191231.
 - Eliminated non-trading days.

Stock index futures

Summary statistics of daily index returns for period 20091231 through 20192131

	S&P 500	RUSS2000	Nasdaq 100	Dow	Midcap 400	DJ RE	S&P Fin
<i>n</i>	2,516	2,516	2,516	2,516	2,516	2,516	2,516
Mean	0.000483	0.000504	0.000711	0.000465	0.000528	0.000365	0.000498
StDev	0.009953	0.013373	0.010964	0.009288	0.011610	0.015936	0.016699
Skewness	-0.315213	-0.135942	-0.245333	-0.298776	-0.283405	0.373483	0.446630
Minimum	-0.068958	-0.093317	-0.063053	-0.057061	-0.086123	-0.115888	-0.121282
Median	0.000362	0.000554	0.000773	0.000374	0.000606	0.000382	0.000273
Maximum	0.068366	0.080660	0.063621	0.066116	0.071101	0.152206	0.163312
CAGR	12.93%	13.54%	19.63%	12.42%	14.24%	9.64%	13.37%
Volatility	15.80%	21.23%	17.41%	14.74%	18.43%	25.30%	26.51%
HPR	236.70%	255.43%	498.47%	221.85%	277.74%	150.62%	249.93%

Nasdaq had highest price appreciation.

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S&P Financial Sector index had highest volatility.

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Major indexes had negative skewness.

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Sector indexes had positive skewness.

Stock index futures

Correlation estimates of daily index returns during period 20091231 through 20192131

	S&P 500	RUSS2000	Nasdaq 100	Dow	Midcap 400	DJ RE	S&P Fin
S&P 500	1						
RUSS2000	0.918	1					
Nasdaq 100	0.925	0.857	1				
Dow	0.975	0.871	0.877	1			
Midcap 400	0.950	0.969	0.877	0.906	1		
DJ RE	0.761	0.750	0.656	0.717	0.780	1	
S&P Fin	0.872	0.817	0.723	0.841	0.840	0.803	1

For major indexes, strongest correlation is between S&P and Dow.

Stock index futures

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For major indexes, strongest correlation is between S&P and Dow.
- Implies Dow futures market will not compete for hedging purposes.

Stock index futures

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For major indexes, weakest correlation is between S&P and Russell 2000.
- Implies Russell 2000 futures market can compete for hedging purposes.

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Stock index futures

- Most active index futures in U.S. is E-mini S&P 500 contract.
 - Is 50 times index level.
 - Expires at open on third Friday of month.
 - Is cash-settled to special index level based on opening trade prices of each of S&P 500 stocks.
 - Weekly expirations have been launched but garner little trading volume.



Composition of stock indexes

- Value-weighted arithmetic indexes
 - S&P 500, Russell 2000, S&P 400, Nasdaq 100
- Price-weighted arithmetic index
 - Dow



Value-weighted index

- S&P 500 index:
 - Included 500 stocks for first time in 1957.
 - Initial divisor was computed using average share prices of index stocks during period 1941-3.
 - Base index level was set equal to 10.
 - Current index level is about 3326.50.
 - Price appreciation of about 33,165%.



Price-weighted index

- DJIA 30:
 - Began with 12 “blue-chip” stocks on May 26, 1896.
 - Average price on that day was 40.94.
 - Increased to 20 stocks in 1916.
 - Increased to 30 stocks in 1928.



Stock index futures valuation

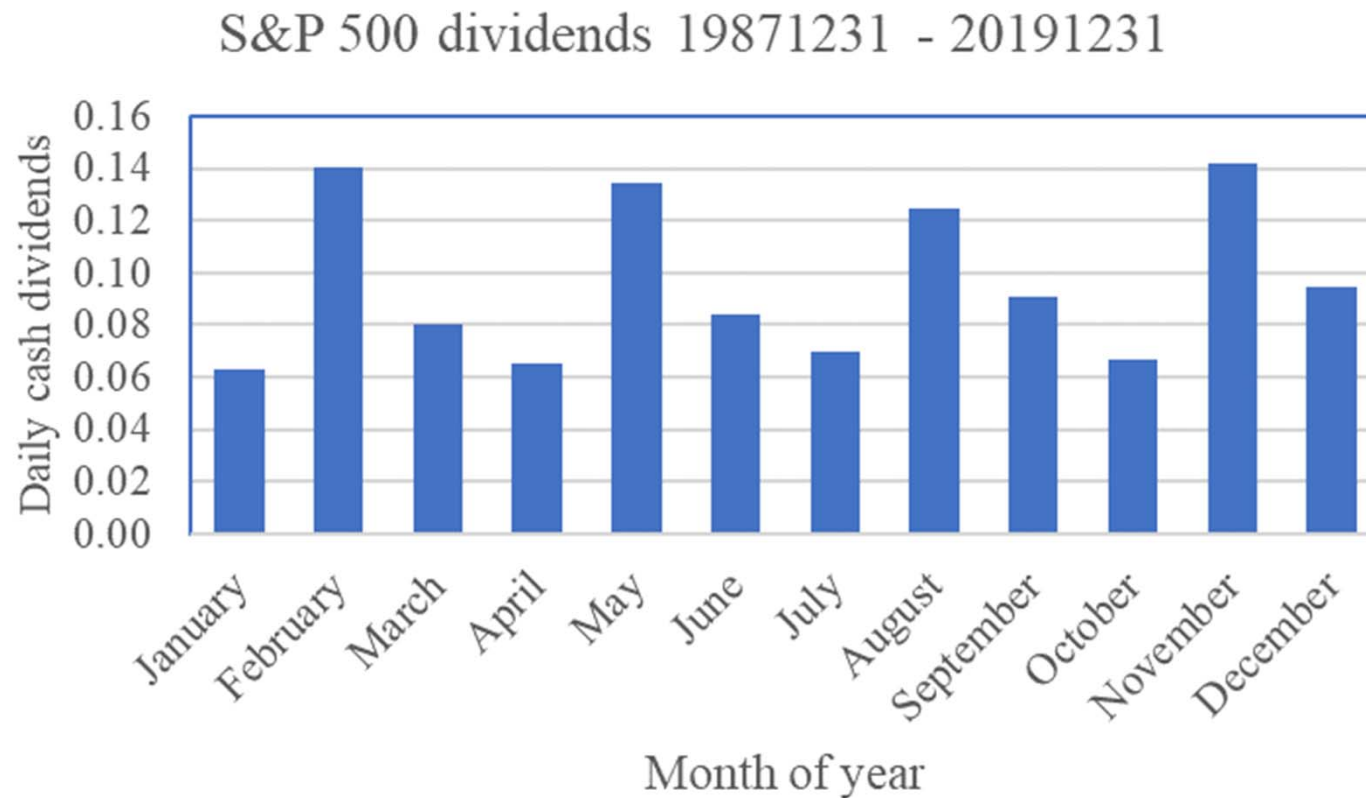
- Earlier, we developed two versions of cost of carry relation.
 - Income is constant continuous rate.
 - Income is discrete payments.
- Which one should be used for stock index futures?



S&P 500 dividends

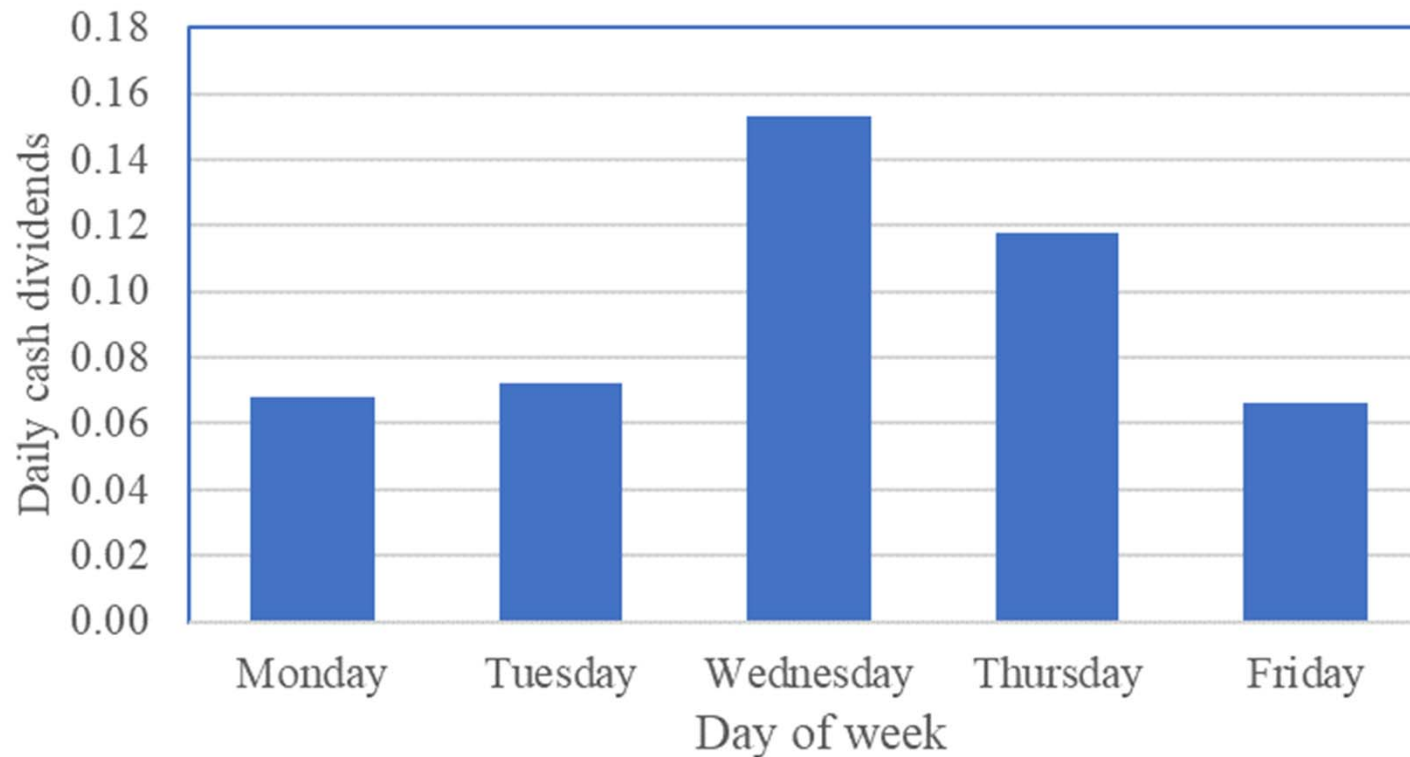
- Supporting file: S&P 500 index dividends.xlsx
 - Downloaded daily data from Datastream for period 19871231 through 20191231.
 - Eliminated non-trading days.

Dividend patterns

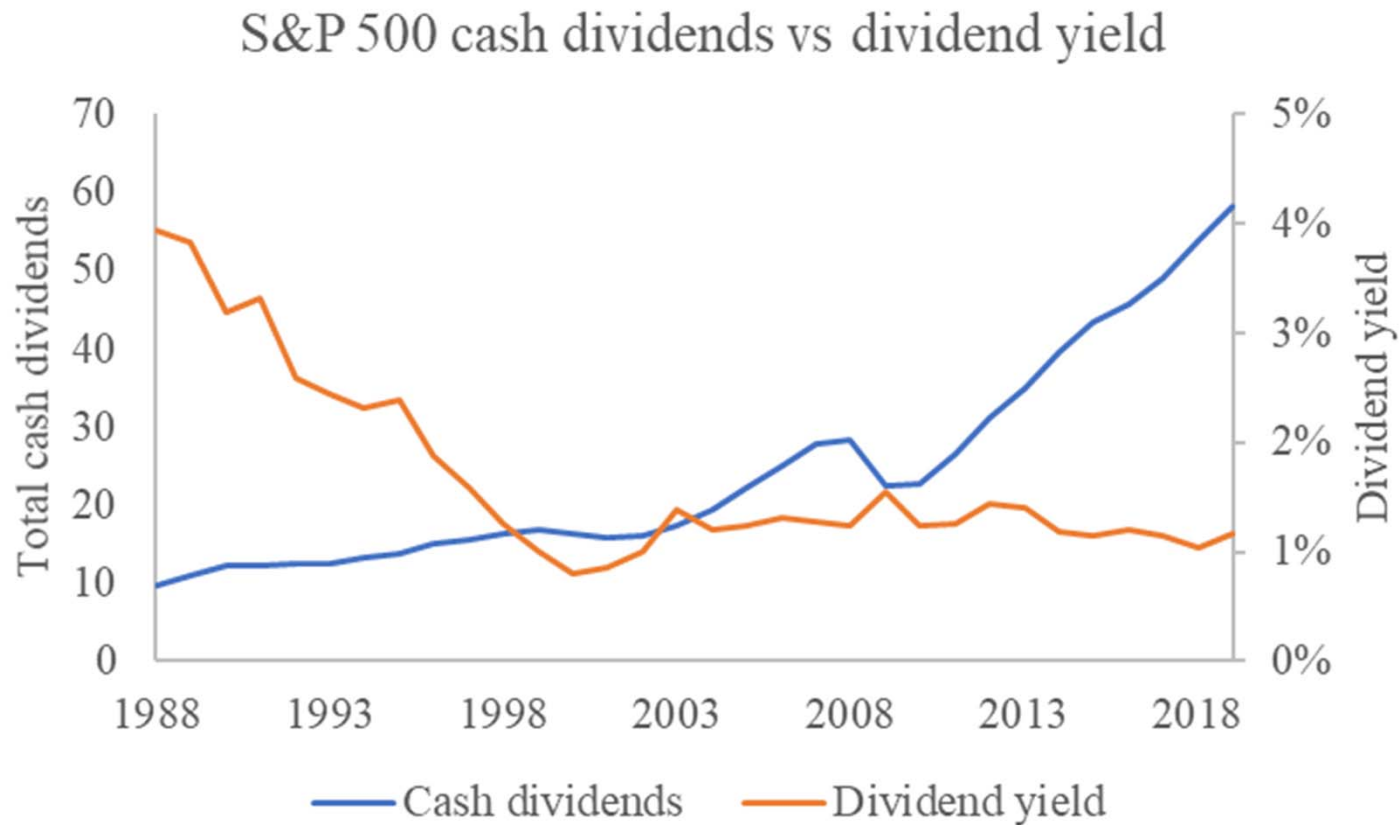


Dividend patterns

S&P 500 dividends 19871231 - 20191231



Dividend patterns





Stock index futures valuation

- Need to handle dividends as discrete flows.
- Net cost of carry relation for stock index futures is

$$F = Se^{rT} - \sum_{i=1}^n D_i e^{r(T-t_i)}$$

Stock index arbitrage

- Trading costs for index arbitrageurs are about 1.50 index points.

- Will execute an arbitrage if

$$F < S e^{rT} - \sum_{i=1}^n D_i e^{r(T-t_i)} - 1.50$$

- Buy futures, sell index portfolio, invest proceeds in T-bills.

Stock index arbitrage

- Trading costs for index arbitrageurs are about 1.50 index points.
 - Will execute an arbitrage if

$$F > Se^{rT} - \sum_{i=1}^n D_i e^{r(T-t_i)} - 1.50$$

- Sell futures, borrow, and buy index portfolio.

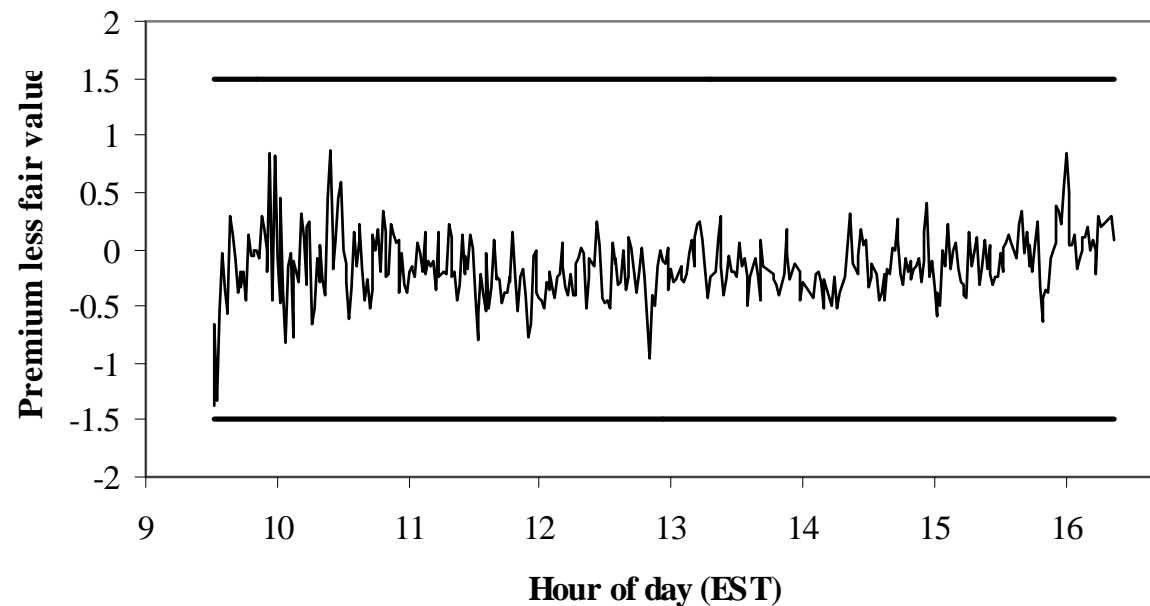
Stock index arbitrage

- How efficient is index arbitrage?
 - On a minute-by-minute basis on 8/29/03, computed

$$\text{Mispricing} = F - \left[S e^{rT} - \sum_{i=1}^n D_i e^{r(T-t_i)} \right]$$

Stock index arbitrage

- How efficient is index arbitrage?
 - Mis-pricing and arbitrage bounds on 8/29/03





Stock index arbitrage

- Buying or selling of index stocks must be *simultaneous*.
 - Requires use of computer-generated orders.
 - Referred to as *basket trading* or *program trading*.



Hedging with index futures

- Manage stock portfolio risk by buying and selling index futures contracts.
 - Assume objective is to minimize risk of hedged portfolio return (subject to given level of return) using available futures contracts.



Hedging with index futures

- Elements of least-risk hedging:
 - Identify least-risk hedge ratio.
 - Show equivalence of hedge ratio to OLS regression slope coefficient.
 - Discuss estimation issues.
 - Generalize model to multiple sources of risk.

Hedging with index futures

□ Notation:

V_0 = initial value of portfolio to be hedged

\tilde{V}_T = uncertain portfolio value at time T

$\tilde{R}_V = \ln\left(\frac{\tilde{V}_T}{V_0}\right)$ = return on portfolio to be hedged

F_0 = initial futures price

\tilde{F}_T = uncertain futures price at time T

$\tilde{R}_F = \ln\left(\frac{\tilde{F}_T}{F_0}\right)$ = return on futures

n_F = number of futures contracts bought (+) or sold (-)

\tilde{R}_H = return on hedged portfolios

Hedging with index futures

- Hedged portfolio return is

$$\tilde{R}_H = \tilde{R}_V + n_F \tilde{R}_F$$

- Risk of hedged portfolio return is

$$\sigma_H^2 = \sigma_V^2 + n_F^2 \sigma_F^2 + 2n_F \sigma_{V,F}$$

- Least-risk hedge is determined by

$$\frac{d\sigma_H^2}{dn_F} = 2n_F^* \sigma_F^2 + 2\sigma_{V,F} = 0$$

Hedging with index futures

- Least-risk hedge ratio is

$$n_F^* = -\frac{\sigma_{V,F}}{\sigma_F^2} = -\rho_{V,F} \left(\frac{\sigma_V}{\sigma_F} \right)$$

- Least-risk hedge ratio if portfolio being hedged underlies futures (e.g., S&P 500)?

$$n_F^* = -\rho_{V,F} \left(\frac{\sigma_V}{\sigma_F} \right) = -1$$



Hedging with index futures

- Illustration: Find least-risk hedge ratio for stock portfolio.
 - Supporting file: Stock portfolio hedge.xlsx
 - Contains daily values of stock portfolio and index stock futures prices.
 - Computes hedge ratio:
 - Analytically
 - Using regression
 - Using SOLVER



Hedging with index futures

- Step 1: Identify appropriate futures contract(s).
 - Since no futures are written on stock portfolio, identify closest, liquid substitute.
 - Examine correlation between stock portfolio and available index futures. Generally want contract with highest correlation.
 - Depth and liquidity of index futures is also important.
 - Depth ensures small price impact.
 - Liquidity (i.e., small bid/ask spread) ensures fast and cost-efficient trading.



Hedging with index futures

- Step 2: Collect historical time series.
 - Daily stock portfolio values
 - Index futures prices
 - Index futures price contract denomination is 50.

Hedging with index futures

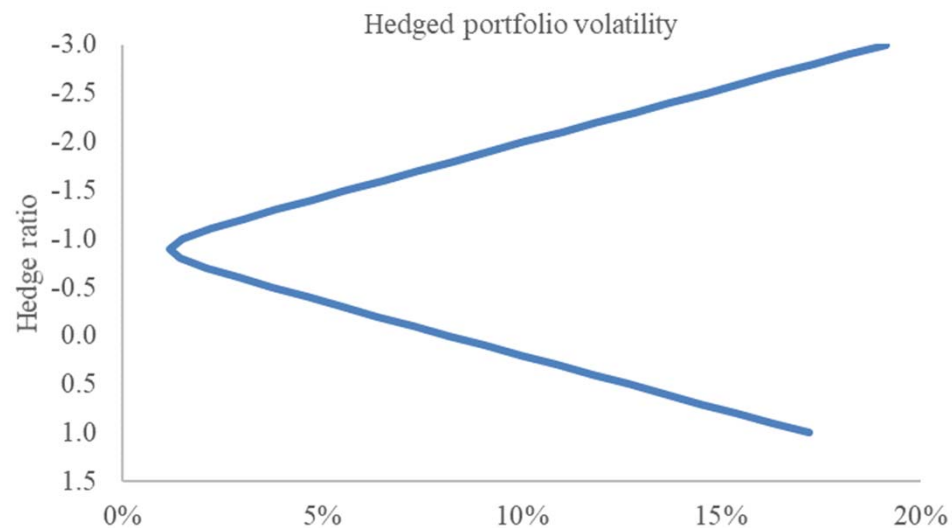
- Step 3: Estimate parameters of minimum risk hedge.
 - Compute:
 - standard deviation of stock portfolio returns
 - standard deviation of futures returns
 - correlation between stock portfolio and futures returns

Summary statistics for returns		
	Stock portfolio return	Index futures return
<i>n</i>	294	294
Mean	0.000405	0.000333
StDev	0.005151	0.005706
Skewness	0.009638	0.075221
Correlation	0.989811	

Hedging with index futures

- Step 4: Compute least-risk hedge ratio.

$$n_F^* = -.989811 \left(\frac{.005151}{.005711} \right) = -.893543$$



Hedging with index futures

- Step 5: Determine least-risk number of futures.
 - Hedge ratio is number of futures to sell per unit of stock portfolio.
 - Need to adjust by dollar value of portfolio and dollar value of stock index futures.

$$\begin{aligned}n_F^* &= -.89354 \left(\frac{32,671,455}{1,152.50 \times 50} \right) \\ &= -.89354 \left(\frac{32,671,455}{1,152.50 \times 50} \right) \\ &= -506.61 \approx -507\end{aligned}$$

Regression approach

- OLS regression provides alternative means for risk measurement.
- Consider regression of portfolio return on futures return.

$$\tilde{R}_V = \alpha_0 + \alpha_1 \tilde{R}_F + \tilde{\varepsilon}$$

- Hedge portfolio return may be written

$$\begin{aligned}\tilde{R}_H &= \tilde{R}_V + n_F \tilde{R}_F \\ &= \alpha_0 + \alpha_1 \tilde{R}_F + \tilde{\varepsilon} + n_F \tilde{R}_F \\ &= \alpha_0 + (\alpha_1 + n_F) \tilde{R}_F + \tilde{\varepsilon}\end{aligned}$$

Regression approach

- Variance of hedged portfolio return is

$$\begin{aligned} \text{Var}(\tilde{R}_H) &= \text{Var}\left[\alpha_0 + (\alpha_1 + n_F)\tilde{R}_F + \tilde{\varepsilon}\right] \\ &= \text{Var}\left[(\alpha_1 + n_F)\tilde{R}_F\right] + \text{Var}(\tilde{\varepsilon}) + 2\text{Cov}\left((\alpha_1 + n_F)\tilde{R}_F, \tilde{\varepsilon}\right) \\ &= \text{Var}\left[(\alpha_1 + n_F)\tilde{R}_F\right] + \text{Var}(\tilde{\varepsilon}) \end{aligned}$$

- Where is hedged portfolio risk minimized?

$$n_F^* = -\alpha_1$$

Regression approach

- Illustration: Find risk-minimizing hedge for portfolio of stocks.
 - Supporting file: Stock portfolio hedge.xlsx

Same risk-minimizing hedge ratio.

Regression Statistics			
Multiple R	0.9898		
R Square	0.9797		
Adjusted R Square	0.9797		
Standard Error	0.0007		
Observations	294		

	<i>Coeff</i>	<i>StErr</i>	<i>t Stat</i>
Intercept	0.00011	0.00004	2.50
Futures return	0.89354	0.00752	118.79

Regression approach

- Illustration: Find risk-minimizing hedge for portfolio of stocks.
 - Supporting file: Stock portfolio hedge.xlsx

Adjusted R Squared is measure of *hedging effectiveness*.

Percent of portfolio return variance explained by futures returns.

Regression Statistics	
Multiple R	0.9898
R Square	0.9797
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Standard Error	0.0007
Observations	294

	<i>Coeff</i>	<i>StErr</i>	<i>t Stat</i>
Intercept	0.00011	0.00004	2.50
Futures return	0.89354	0.00752	118.79



Estimation issues

- Return interval must be selected (e.g., daily, weekly, monthly, etc. returns)
 - higher frequency implies more information (good) but also more noise (bad)
 - prices for cash and futures must be simultaneous



Estimation issues

- Illustration: Suppose we hold S&P 500 index portfolio and want to hedge market risk using S&P 500 futures.
 - Already know least-risk hedge ratio is -1 .



Estimation issues

- Suppose we use historical data to estimate hedge.
- S&P 500 futures data during 1997
 - 1997 had 254 trading days, which creates
 - 253 daily returns
 - 52 weekly returns
 - 26 biweekly returns
 - For weekly and biweekly returns, Wednesday closing prices are used.



Estimation issues

- Use returns of nearby futures contract.
 - When switching contract months, care must be taken to splice price change series correctly.
- Supporting file: S&P 500 hedge.xlsx

Estimation issues

- Daily, weekly, and bi-weekly regressions using S&P 500 futures data during 1997.

Hedge regressions using different return intervals

	α_0	$t(\alpha_0)$	α_1	$t(\alpha_1)$	Adjusted R-squared
Daily	0.0003	1.65	0.8764	70.07	0.9514
Weekly	0.0009	2.29	0.9558	55.64	0.9841
Bi-weekly	0.0016	2.61	0.9884	43.40	0.9874

Estimation issues

- Daily, weekly, and bi-weekly regressions using S&P 500 futures data during 1997.

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Bi-weekly	0.0016	2.61	0.9884	43.40	0.9874

Measurement error can be large.
Correct answer is 1.



Estimation issues

- CAVEAT: Parameters are estimated using past data, but we are interested in future.
 - Must have reason to believe relation is stationary.



Hedging multiple risks

- OLS approach can be generalized to handle multiple sources of risk.

$$\tilde{R}_V = \alpha_0 + \alpha_1 \tilde{R}_{F,1} + \alpha_2 \tilde{R}_{F,2} + \dots + \alpha_n \tilde{R}_{F,n} + \tilde{\varepsilon}$$

- Set hedge ratios for all or just selected risks.



Hedging multiple risks

- Illustration: Hedge oil price risk of fund that invests primarily in oil stocks.



Hedging multiple risks

- Step 1: Identify portfolio risk exposures and find futures to proxy for each.
 - SP: S&P 500 futures (stock market risk)
 - CL: crude oil futures (petroleum price risk)
 - HO: heating oil futures (processed petroleum price risk)
 - HU: unleaded gas futures (processed petroleum price risk)

Hedging multiple risks

- Step 2: Collect historical return data.
 - Supporting file: Oil hedge.xlsx

Day (\$ millions)	Portfolio		Futures prices				Futures returns			
	value	Portfolio return	SP	CL	HO	HU	SP	CL	HO	HU
1	769.26		1100.00	40.00	0.6900	0.7500				
2	776.57	0.00946	1099.00	40.39	0.6969	0.7616	-0.00091	0.00970	0.00995	0.01535
3	782.45	0.00754	1100.55	40.50	0.7051	0.7709	0.00141	0.00272	0.01170	0.01214
4	791.62	0.01165	1101.10	41.17	0.7236	0.7878	0.00050	0.01641	0.02590	0.02169
5	791.44	-0.00023	1100.90	41.25	0.7213	0.7865	-0.00018	0.00194	-0.00318	-0.00165
6	789.32	-0.00268	1104.30	41.15	0.7265	0.7882	0.00308	-0.00243	0.00718	0.00216
7	793.38	0.00513	1109.05	40.50	0.7124	0.7670	0.00429	-0.01592	-0.01960	-0.02727
8	789.23	-0.00524	1108.40	40.68	0.7290	0.7764	-0.00059	0.00443	0.02303	0.01218
9	786.04	-0.00405	1107.25	40.16	0.7255	0.7673	-0.00104	-0.01287	-0.00481	-0.01179
10	784.78	-0.00160	1106.65	40.34	0.7339	0.7690	-0.00054	0.00447	0.01151	0.00221

Hedging multiple risks

- Step 3: Get to know properties of data.

SP volatility low
during sample period.

Summary statistics				
	Futures returns			
	SP	CL	HO	HU
<i>n</i>	252	252	252	252
Mean	-0.000029	0.000483	0.000773	0.000609
StDev	0.002798	0.006899	0.011017	0.011879
Skewness	-0.473383	-0.162212	-0.207878	-0.265877
CAGR	-0.73%	12.95%	21.49%	16.60%
Volatility	4.44%	10.95%	17.49%	18.86%

Correlations				
	SP	CL	HO	HU
SP	1			
CL	-0.191	1		
HO	-0.135	0.773	1	
HU	-0.079	0.718	0.662	1

Hedging multiple risks

- Step 3: Get to know properties of data.

SP volatility low during sample period.

CL and processed products are more normal.

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Hedging multiple risks

- Step 3: Get to know properties of data.

SP volatility low during sample period.

CL and processed products are more normal.

CL less volatility than HO or HU.

Summary statistics				
	Futures returns			
	SP	CL	HO	HU
<i>n</i>	252	252	252	252
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Hedging multiple risks

- Step 3: Get to know properties of data.

What does this say about oil
oil stock exposure?

Summary statistics				
	Futures returns			
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Hedging multiple risks

- Step 3: Get to know properties of data.

		Summary statistics			
		Futures returns			
		SP	CL	HO	HU
	<i>n</i>	252	252	252	252
What does this say about oil oil stock exposure?	Mean	-0.000029	0.000483	0.000773	0.000609
	StDev	0.002798	0.006899	0.011017	0.011879
	Skewness	-0.473383	-0.162212	-0.207878	-0.265877
Are stocks:	CAGR	-0.73%	12.95%	21.49%	16.60%
1) Oil exploration	Volatility	4.44%	10.95%	17.49%	18.86%
2) Oil refining, or					
3) Oil production distribution?					
		Correlations			
		SP	CL	HO	HU
SP		1			
CL		-0.191	1		
HO		-0.135	0.773	1	
HU		-0.079	0.718	0.662	1

Hedging multiple risks

- Step 4: Run OLS regression.

$$\tilde{R}_V = \alpha_0 + \alpha_1 \tilde{R}_{F,1} + \alpha_2 \tilde{R}_{F,2} + \dots + \alpha_n \tilde{R}_{F,n} + \tilde{\varepsilon}$$

Regression Statistics	
Multiple R	0.5826
R Square	0.3394
Adjusted R Square	0.3287
Standard Error	0.0060
Observations	252

	<i>Coeff</i>	<i>StErr</i>	<i>t Stat</i>
Intercept	0.00002	0.00038	0.05
SP	1.55502	0.13904	11.18
CL	0.21055	0.09773	2.15
HO	-0.03085	0.05610	-0.55
HU	-0.00062	0.04764	-0.01

Hedging multiple risks

- Step 4: Run OLS regression.

$$\tilde{R}_V = \alpha_0 + \alpha_1 \tilde{R}_{F,1} + \alpha_2 \tilde{R}_{F,2} + \dots + \alpha_n \tilde{R}_{F,n} + \tilde{\varepsilon}$$

Hedge is not (will not be) particularly effective.

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Hedging multiple risks

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HO and HU returns have little effect on portfolio return.

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Hedging multiple risks

- Step 4: Run OLS regression.

$$\tilde{R}_V = \alpha_0 + \alpha_1 \tilde{R}_{F,1} + \alpha_2 \tilde{R}_{F,2} + \dots + \alpha_n \tilde{R}_{F,n} + \tilde{\varepsilon}$$

HO and HU returns have little effect on portfolio return.

Drop HO and HU and re-run regression.

Regression Statistics			
Multiple R	0.5826		
R Square	0.3394		
Adjusted R Square	0.3287		
Standard Error	0.0060		
Observations	252		

	<i>Coeff</i>	<i>StErr</i>	<i>t Stat</i>
Intercept	0.00002	0.00038	0.05
SP	1.55502	0.13904	11.18
CL	0.21055	0.09773	2.15
HO	-0.03085	0.05610	-0.55
HU	-0.00062	0.04764	-0.01

Hedging multiple risks

- Step 5: Drop heating oil and crude oil and re-run regression.

Adjusted R-squared is increased.

Regression Statistics	
Multiple R	0.5818
R Square	0.3385
Adjusted R Square	0.3332
Standard Error	0.0060
Observations	252

	<i>Coeff</i>	<i>StErr</i>	<i>t Stat</i>
Intercept	0.00001	0.00038	0.04
SP	1.55328	0.13808	11.25
CL	0.17158	0.05600	3.06

Hedging multiple risks

- Step 5: Drop heating oil and crude oil and re-run regression.

Adjusted R-squared is increased.

Both SP and CL returns explain portfolio returns.

Regression Statistics			
Multiple R	0.5818		
R Square	0.3385		
Adjusted R Square	0.3332		
Standard Error	0.0060		
Observations	252		

	<i>Coeff</i>	<i>StErr</i>	<i>t Stat</i>
Intercept	0.00001	0.00038	0.04
SP	1.55328	0.13808	11.25
CL	0.17158	0.05600	3.06



Hedging multiple risks

- Step 6: Find number of crude oil futures to sell.

$$n_F^* = -.17158 \left(\frac{779,520,000}{45.18 \times 1,000} \right) = -2,960.36 \approx -2,960$$

Estimation issues

- Missing variable bias: Regress on crude oil futures return only.

Regression Statistics			
Multiple R	0.0482		
R Square	0.0023		
Adjusted R Square	-0.0017		
Standard Error	0.0074		
Observations	252		

	<i>Coeff</i>	<i>StErr</i>	<i>t Stat</i>
Intercept	0.00003	0.00047	0.06
CL	0.05136	0.06738	0.76

$$n_F^* = -.05136 \left(\frac{779,520,000}{45.18 \times 1,000} \right) = -886.22 \approx -886$$

Estimation issues

- Missing variable bias: Recall return correlations.

Correlations				
	SP	CL	HO	HU
SP	1			
CL	-0.191	1		
HO	-0.135	0.773	1	
HU	-0.079	0.718	0.662	1

Absence of SP means that CL, to some degree, also picks up effect of CL because SP and CL are correlated.

Negative correlation implies downward bias.



Lesson summary

- Stock index futures were introduced in U.S. in 1982; stock index options in 1983.
- Construction of stock indexes
 - value-weighted indexes
 - price-weighted indexes
- Cash dividends
 - use discrete cash flows in valuation



Lesson summary

- Index arbitrage
 - Program trading
- *OLS regression* of portfolio returns on futures returns provides estimate of minimum-risk hedge ratio.
- *Multiple regression* can be used to estimate hedge ratios for portfolios with multiple sources of risk.



Lesson summary

- Steps in setting multiple risk portfolio hedge:
 - Identify portfolio risk exposures and find futures to proxy for each.
 - Collect historical return series.
 - Estimate OLS regression.
 - Use coefficient estimates to hedge selected exposures.



Lesson summary

- Careful data analysis is necessary to properly estimate hedge ratios. Examined effects of:
 - Data frequency
 - Missing variables bias
 - Failing to include relevant explanatory variable is much more serious than including irrelevant explanatory variable.