

Lecture 5: Mean-Reversion

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Stationarity/ Non Stationarity

Definition: a stochastic process is stationary if

$$\forall m, \forall (t_1, \dots, t_m), \forall A \in \mathbf{R}^n$$
$$\Pr.\{(X_{t_1}, X_{t_2}, \dots, X_{t_m}) \in A\} = \Pr.\{(X_{t_1+h}, X_{t_2+h}, \dots, X_{t_m+h}) \in A\}$$

A stationary process is a process that is “statistically invariant under translations”

Examples: the Ornstein-Uhlenbeck process is stationary, Brownian motion is not.

The Ornstein-Uhlenbeck process

$$dX_t = \kappa(m - X_t)dt + \sigma dW_t$$

$$X_t = e^{-\kappa(t-s)} X_s + (1 - e^{-\kappa(t-s)})m + \sigma \int_s^t e^{-\kappa(t-u)} dW_u$$

$$X_t = m + \sigma \int_{-\infty}^t e^{-\kappa(t-s)} \eta(s) ds, \quad \eta(s) = \text{Gaussian white noise}$$

Exponentially-weighted moving average of uncorrelated Gaussian random variables.

Statistics of the OU process

$$\begin{aligned}\langle X_t X_{t+h} \rangle &= \sigma^2 \left\langle \int_{-\infty}^t e^{-k(t-s)} \eta(s) ds \cdot \int_{-\infty}^{t+h} e^{-k(t+h-s')} \eta(s') ds' \right\rangle \\ &= \sigma^2 \int_{-\infty}^t \int_{-\infty}^{t+h} e^{-k(t-s)} e^{-k(t+h-s')} \delta(s-s') ds ds' \\ &= \sigma^2 \int_{-\infty}^t e^{-k(t-s)} e^{-k(t+h-s)} ds \\ &= \sigma^2 e^{-kh} \int_{-\infty}^t e^{-2k(t-s)} ds \\ &= \frac{\sigma^2 e^{-kh}}{2k}\end{aligned}$$

$$\langle |X_{t+h} - X_t|^2 \rangle = \frac{\sigma^2}{k} (1 - e^{-kh})$$

Structure Function

Random Walk, Fractional BM

$$X_t = \sigma W_t, \quad W_t = \text{Brownian motion}$$

$$\langle |X_{t+h} - X_t|^2 \rangle = \sigma^2 h \quad \langle X_{t+h} X_t \rangle = t$$

Brownian motion (non-stationary)
Structure fn grows linearly

$$X_t = \sigma \int_{-\infty}^t \frac{\eta(s) ds}{(1+t-s)^p} \quad p > 1/2$$

Geometric Brownian motion

$$\langle X_t X_{t+h} \rangle = \frac{\sigma^2}{h^{2p-1}} \int_{\frac{1}{h}}^{\infty} \frac{du}{u^p (1+u)^p}$$

$$\langle X_t X_{t+h} \rangle \approx \left\{ \begin{array}{ll} \frac{\sigma^2}{h^{2p-1}} & 1/2 < p < 1 \\ \frac{\sigma^2 \ln(h)}{h} & p = 1 \\ \frac{\sigma^2}{h^p} & p > 1 \end{array} \right.$$

Correlations decay like power-laws
(large h)

Autoregressive Models

$$X_1, X_2, \dots, X_n, \dots$$

$$X_{n+1} = a_0 + a_1 X_n + \dots + a_m X_{n-m+1} + \sigma v_{n+1}, \quad v_i \sim N(0,1)$$

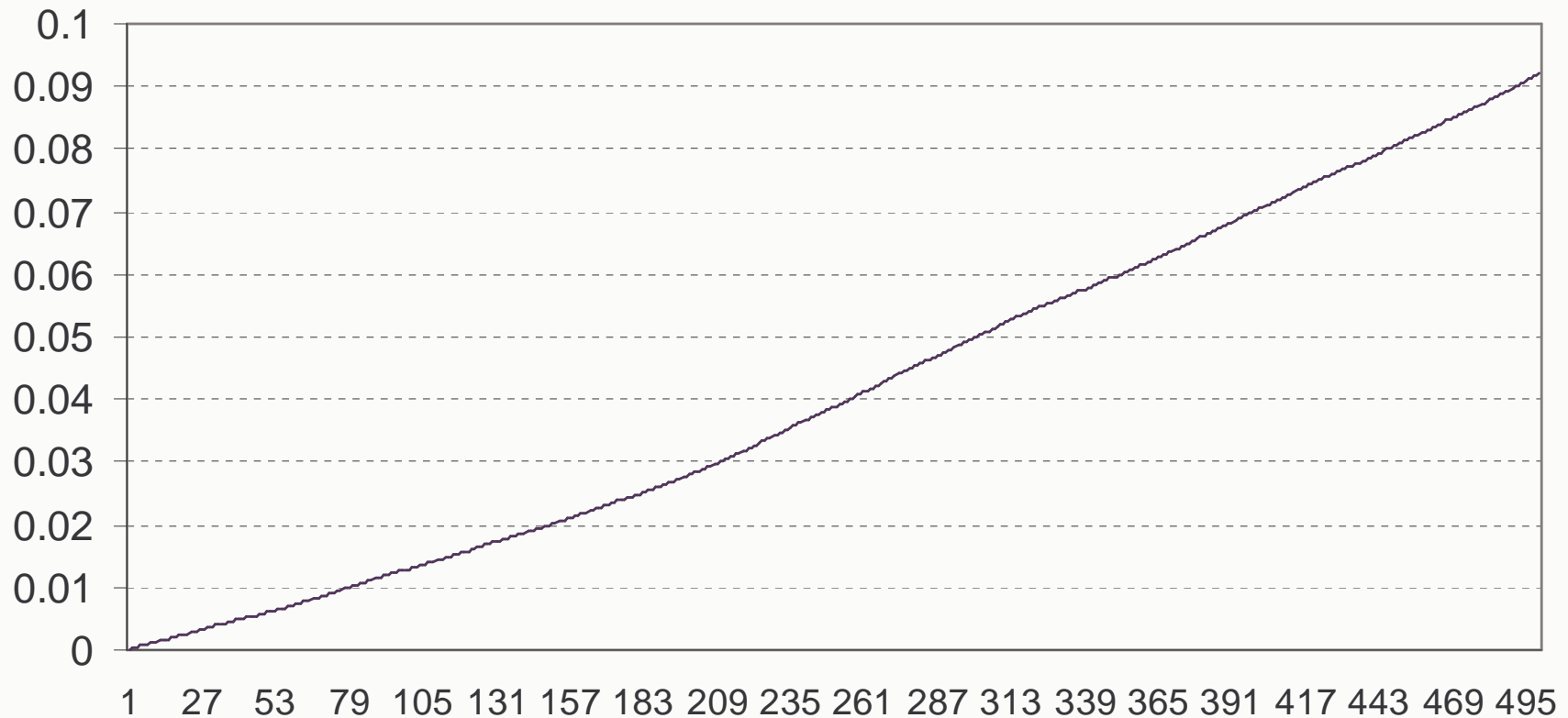
$$\mathbf{Y}_n = \begin{pmatrix} X_n \\ \dots \\ X_{n-m+1} \end{pmatrix}, \quad \mathbf{A} = \begin{pmatrix} a_1 & \dots & a_m \\ 1 & 0 & \dots \\ \dots & 1 & 0 \end{pmatrix}, \quad \mathbf{A}_0 = \begin{pmatrix} a_0 \\ 0 \\ \dots \end{pmatrix}, \quad \mathbf{\Sigma} = \begin{pmatrix} \sigma \\ 0 \\ \dots \end{pmatrix}$$

$$\mathbf{Y}_{n+1} = \mathbf{A}_0 + \mathbf{A}\mathbf{Y}_n + \mathbf{\Sigma}v_{n+1}$$

AR-n model corresponds to a vector AR-1 model

Structure function: SPY Jan 1996-Jan 2009

Use log prices as time series. Structure function with lags 1 day to 2 yrs

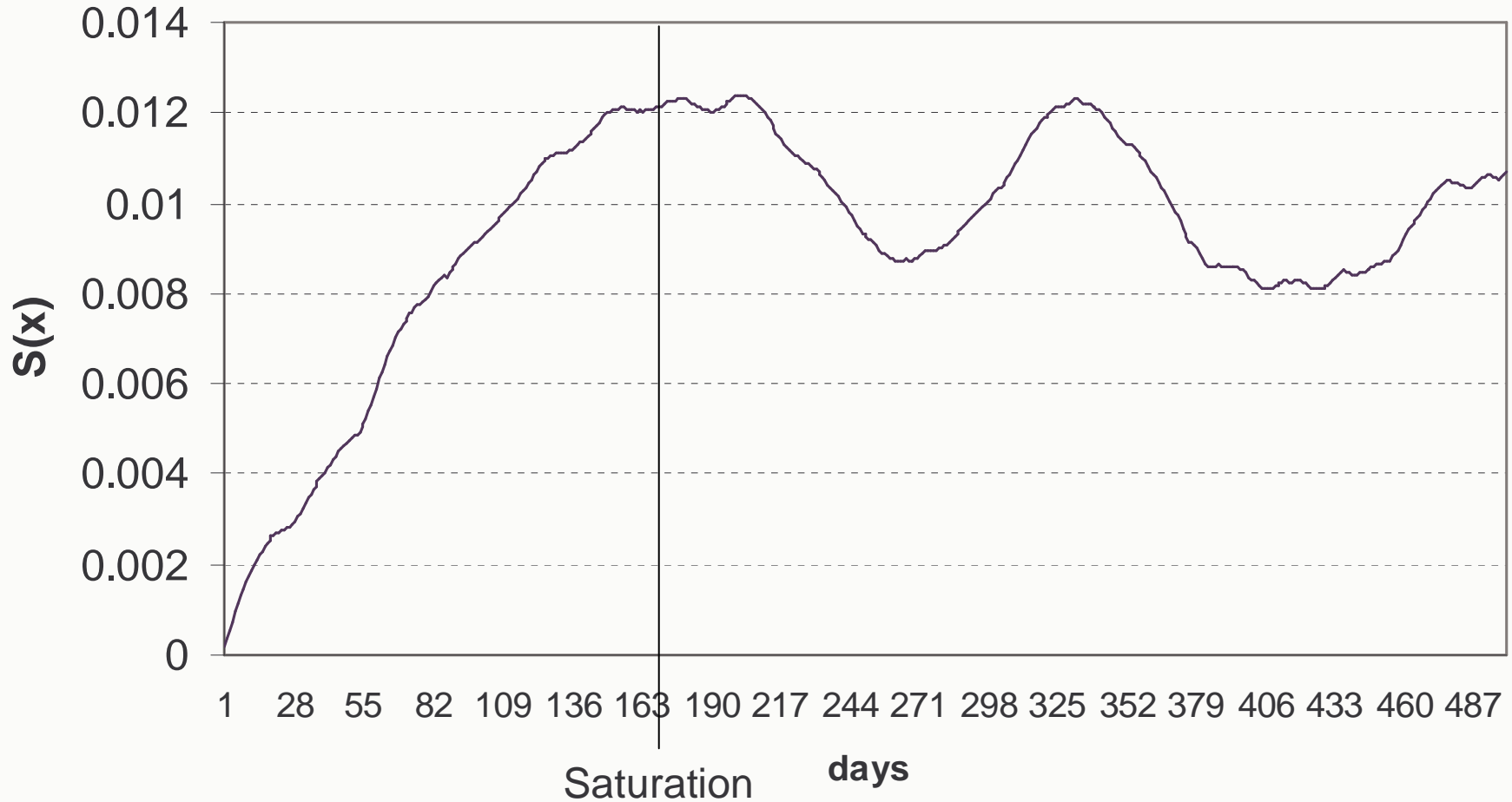


SPY is highly non stationary, as shown in the chart.

Look for mean-reversion in relative value, i.e. in terms of two or more assets.

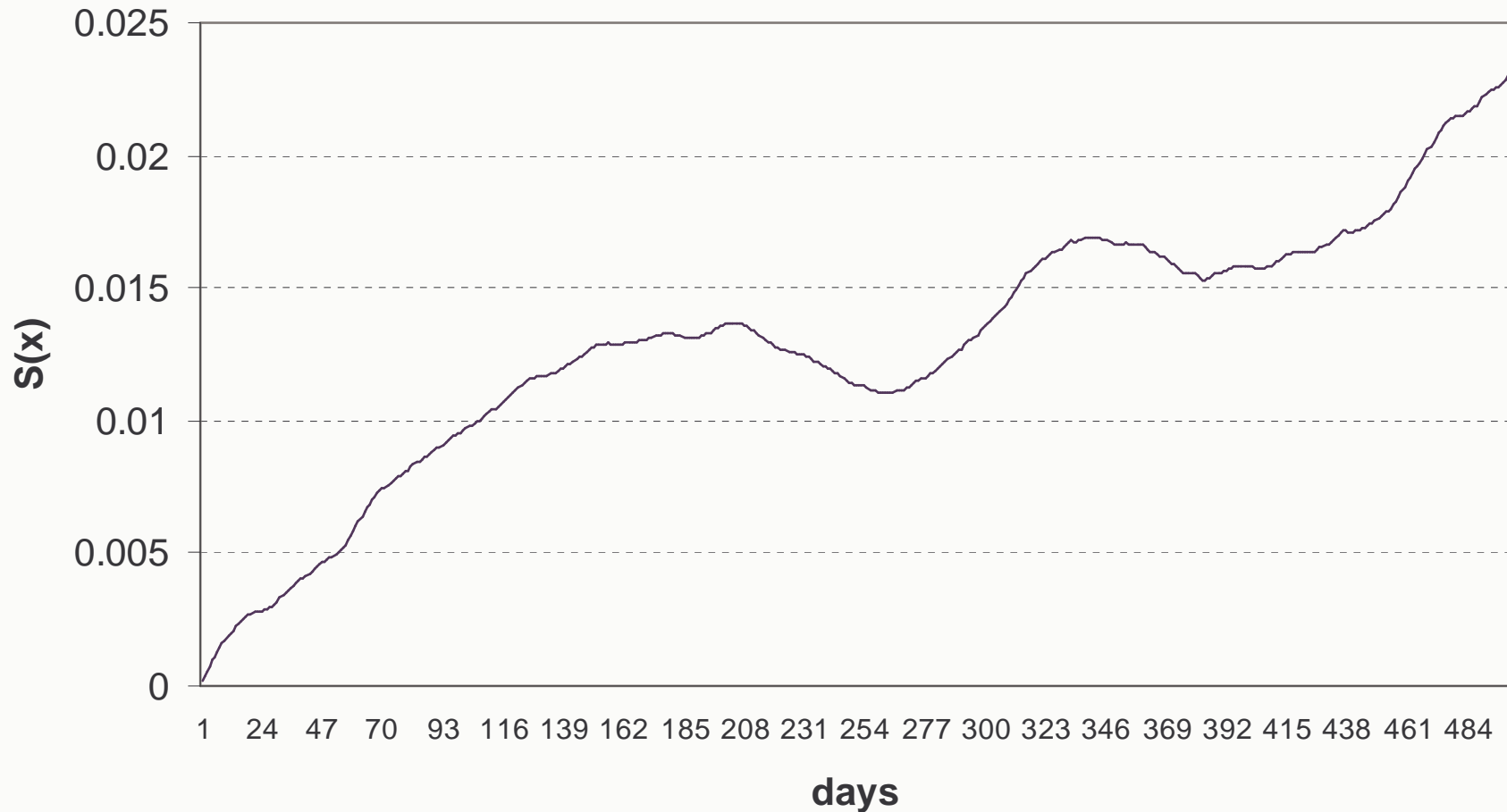
Structure function log (SLB/OIH)

Data: Apr 2006 to Feb 2009



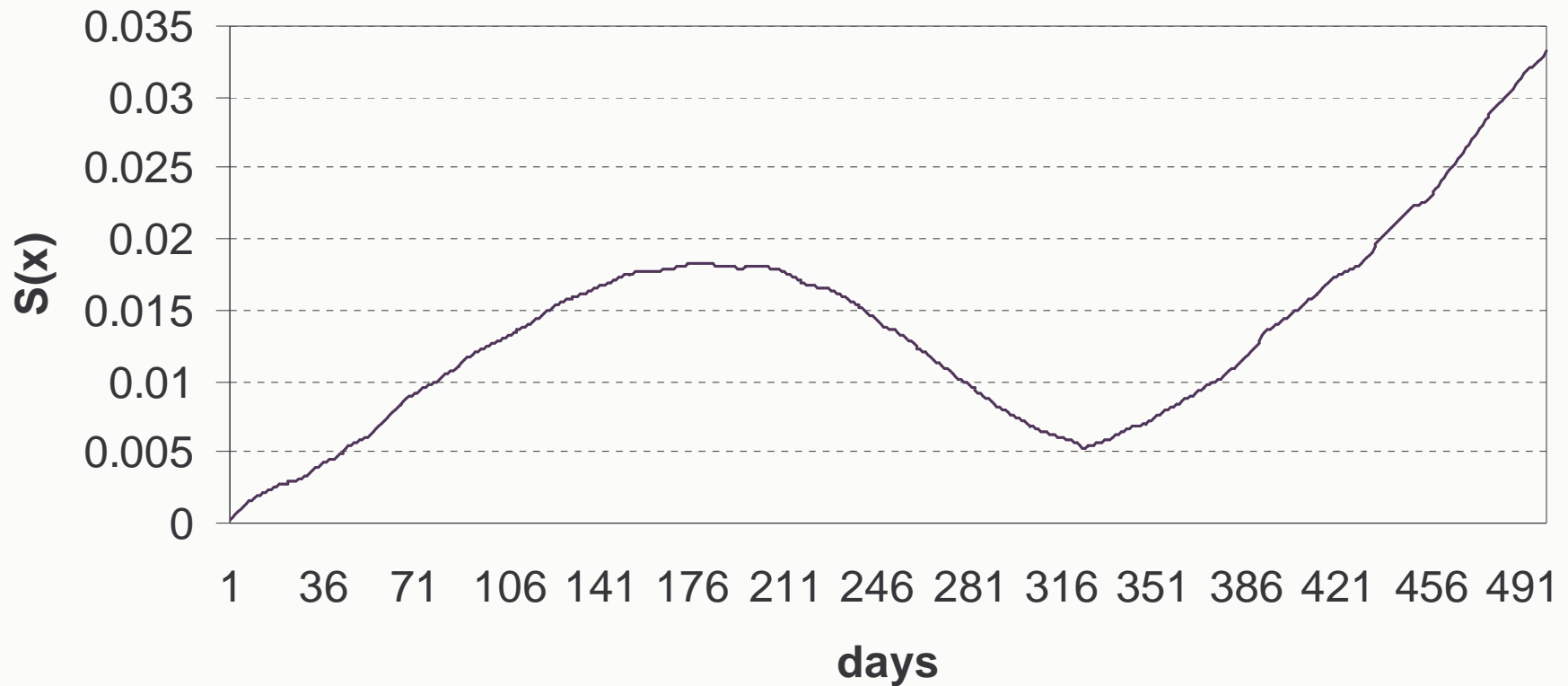
OIH: Oil Services ETF, SLB: Schlumberger-Doll Research

Structure Function: long-short equal dollar weighted SLB-OIH



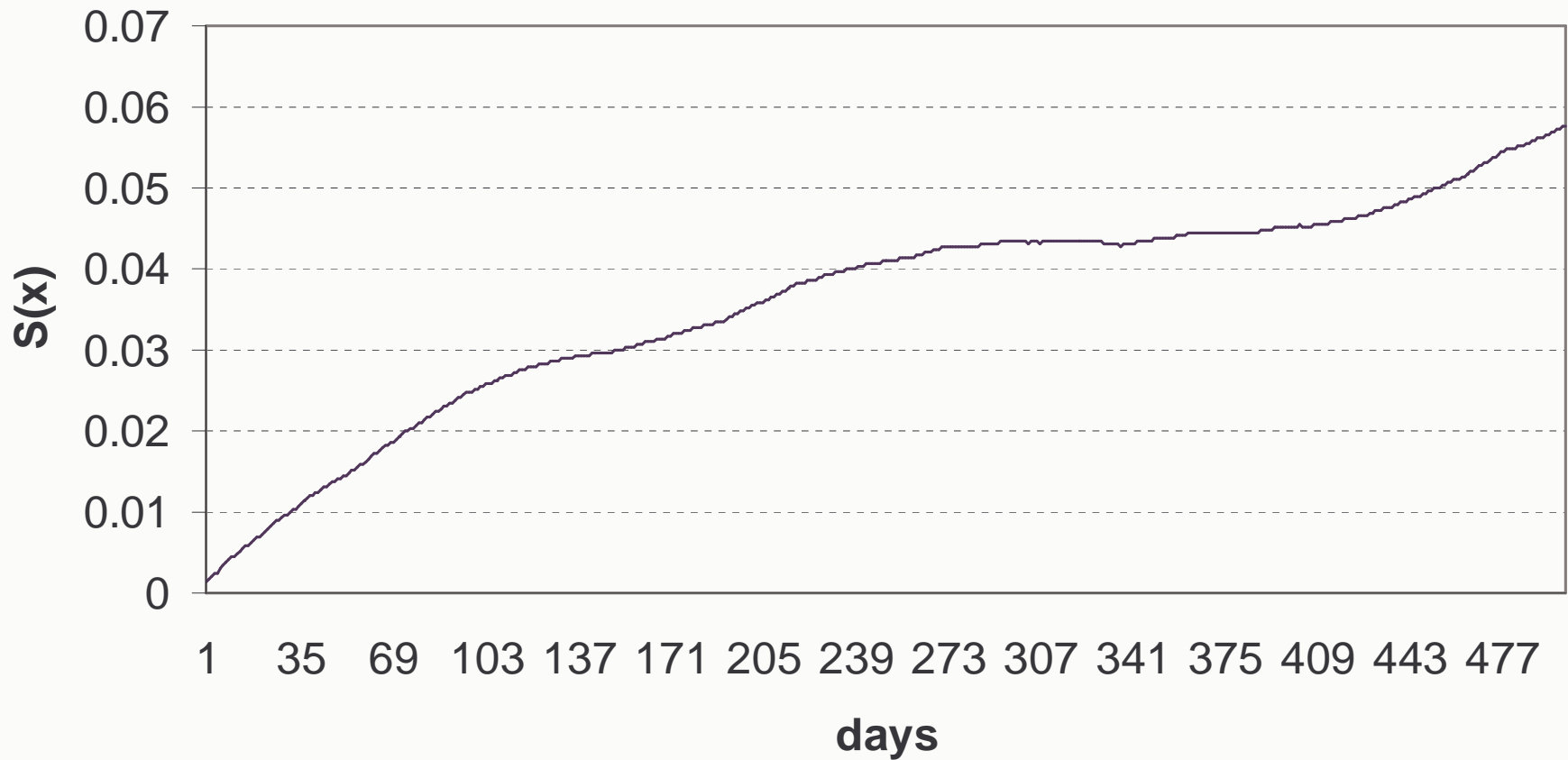
$$P_{n+1} = P_n \times (1 + R_{\text{slb}} - R_{\text{oih}}), \quad X_n = \ln P_n$$

Structure Function for Beta-Neutral long-short portfolio SLB-Beta*OIH

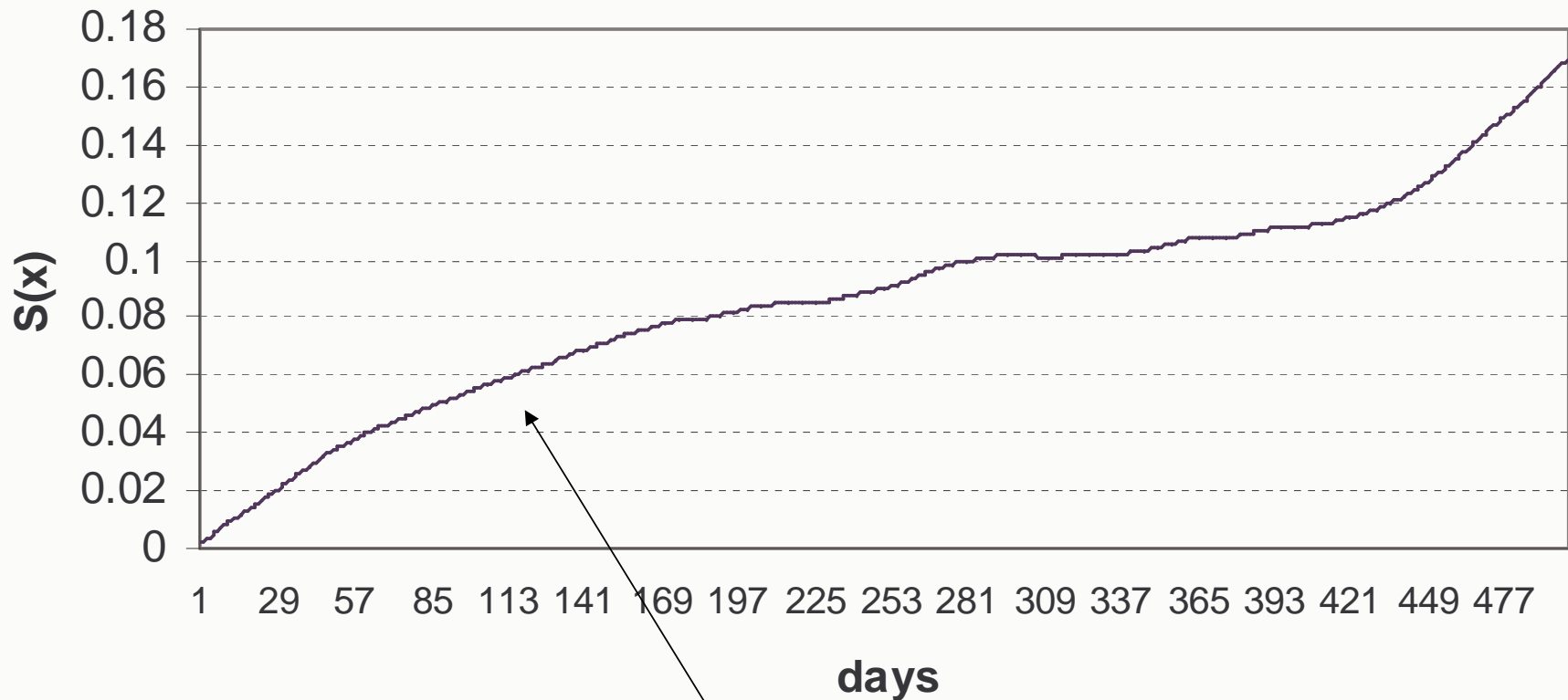


$$P_{n+1} = P_n \times (1 + R_{slb} - \beta_{60d} \cdot R_{oih}), \quad X_n = \ln P_n$$

Structure Function log (GENZ/IBB)



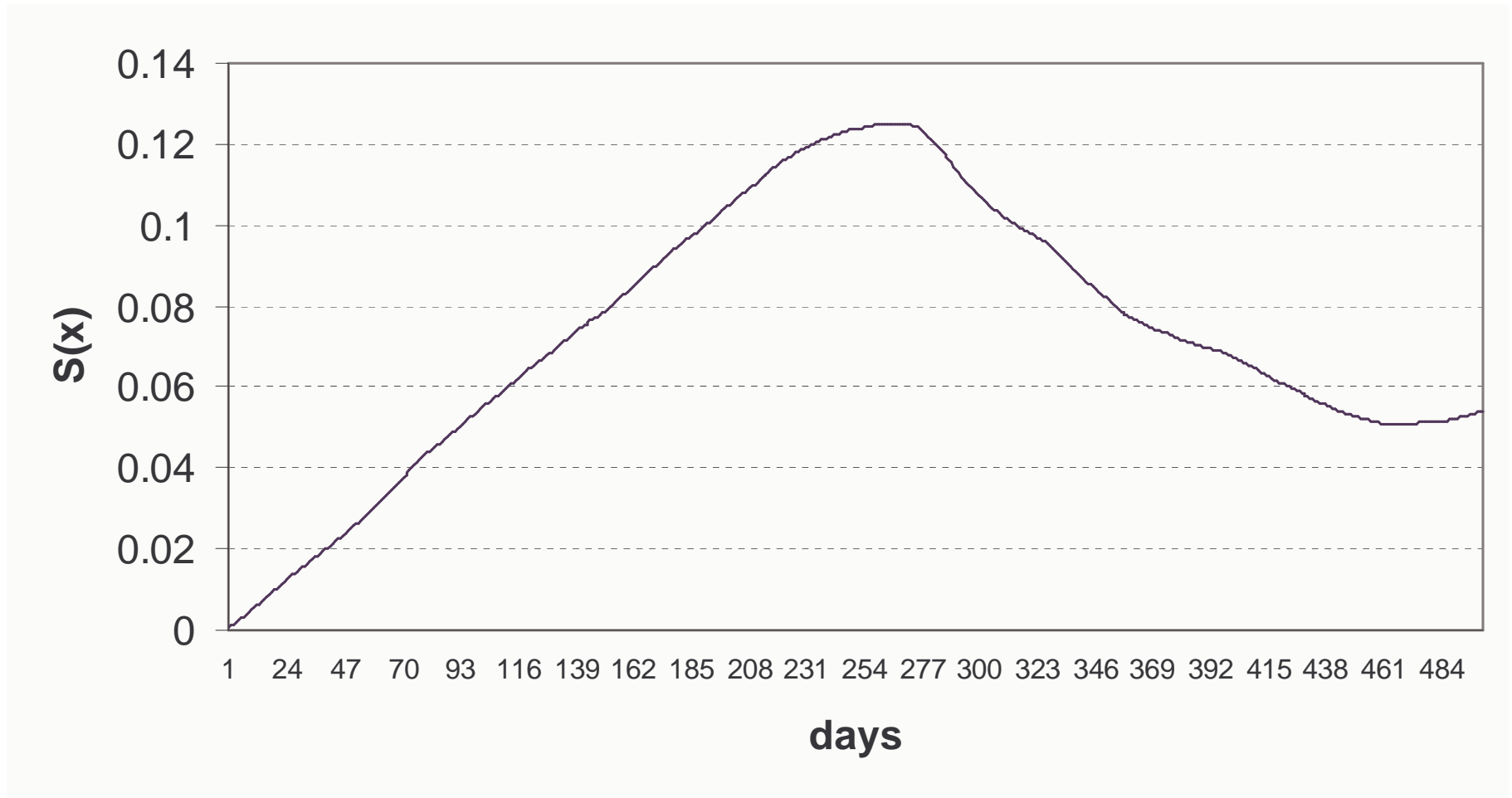
Structure function $\ln(DNA/GENZ)$



DNA: Genentech Inc.
GENZ; Genzyme Corp.

Mean-reversion: large negative curvature here.

Structure Fn for Beta-Neutral GENZ-DNA Spread



Poor reversion for the beta adjusted pair. Beta is low ~ 0.30

Systematic Approach for looking for MR in Equities

Look for stock returns devoid of explanatory factors, and analyze the corresponding residuals as stochastic processes.

$$R_t = \sum_{k=1}^m \beta_k F_{kt} + \varepsilon_t$$

Econometric factor model

$$X_t = X_0 + \sum_{s=1}^t \varepsilon_s$$

View residuals as increments of a process that will be estimated

$$\frac{dS(t)}{S(t)} = \sum_{k=1}^m \beta_k \frac{dP_k(t)}{P_k(t)} + dX(t)$$

Continuous-time model for evolution of stock price

Interpretation of the model

The factors are either

- A. eigenportfolios corresponding to significant eigenvalues of the market
- B. industry ETF, or portfolios of ETFs

Questions of interest:

Can residuals be fitted to (increments of) OU processes or other MR processes?

If so, what is the typical correlation time-scale?

Estimation of Ornstein-Uhlenbeck models

$$X_{t+\Delta t} = e^{-k\Delta t} X_t + m(1 - e^{-k\Delta t}) + \sigma \int_t^{t+\Delta t} e^{-k(t-s)} dW_s$$

$$X_{n+1} = aX_n + b + v_{n+1} \quad \{v_n\} \text{ i.i.d. } N\left(0, \sigma^2 \left(\frac{1 - e^{-2k\Delta t}}{2k}\right)\right)$$

$$a = \text{SLOPE}((X_{n-l}, \dots, X_n); (X_{n-l-1}, \dots, X_{n-1})),$$

$$b = \text{INTERCEPT}((X_{n-l}, \dots, X_n); (X_{n-l-1}, \dots, X_{n-1}))$$

$$k = \frac{1}{\Delta t} \ln\left(\frac{1}{a}\right), \quad m = \frac{b}{1-a}, \quad \sigma = \frac{\text{STDEV}(X_{n+1} - aX_n - b)}{\sqrt{1-a^2}} \sqrt{2 \frac{1}{\Delta t} \ln\left(\frac{1}{a}\right)}$$

Portfolio Strategy

Q_1, Q_2, \dots, Q_N \$ invested in different stocks (long or short)

S_1, S_2, \dots, S_N dividend - adjusted prices

$$\begin{aligned}d\Pi &= \sum_{i=1}^N Q_i \frac{dS_i}{S_i} - \left(\sum_{i=1}^N Q_i \right) r dt \quad (\text{neglect transaction costs}) \\ &= \sum_{i=1}^N Q_i \left(\sum_{k=1}^m \beta_{ik} \frac{dP_k}{P_k} + dX_i \right) - \left(\sum_{i=1}^N Q_i \right) r dt \\ &= \sum_{i=1}^N Q_i dX_i + \sum_{k=1}^m \left(\sum_{i=1}^N Q_i \beta_{ik} \right) \frac{dP_k}{P_k} - \left(\sum_{i=1}^N Q_i \right) r dt\end{aligned}$$

$\sum_{i=1}^N Q_i \beta_{ik}$: net dollar - beta exposure along factor k

$\left(\sum_{i=1}^N Q_i \right)$: net dollar exposure of portfolio

Market-Neutral Portfolio

Assume $dX_i = k_i(m - X_i)dt + \sigma_i dW_i$ $\{dW_i\}_{i=1}^N$ uncorrelated

$$\begin{aligned}d\Pi &= \sum_{i=1}^N Q_i dX_i - \left(\sum_{i=1}^N Q_i \right) r dt \\ &= \sum_{i=1}^N Q_i (k_i(m - X_i)dt + \sigma_i dW_i) - \left(\sum_{i=1}^N Q_i \right) r dt \\ &= \sum_{i=1}^N Q_i (k_i(m - X_i) - r) dt + \sum_{i=1}^N Q_i \sigma_i dW_i\end{aligned}$$

\therefore

$$E(d\Pi | \mathbf{X}) = \sum_{i=1}^N Q_i (k_i(m - X_i) - r) dt$$

$$\text{Var}(d\Pi | \mathbf{X}) = \sum_{i=1}^N Q_i^2 \sigma_i^2 dt$$

Mean-Variance Optimal Portfolio

$$\max_Q \left(\sum_i Q_i \mu_i - \frac{1}{2\lambda} \sum_i Q_i^2 \sigma_i^2 \right) \quad \therefore \quad Q_i = \lambda \frac{\mu_i}{\sigma_i^2}$$

(if $r = 0$, or $\sum Q_i = 0$)

$$d\Pi = \lambda \sum_i \frac{k_i^2 (m - X_i)^2}{\sigma_i^2} dt + \lambda \sum_i \frac{k_i (m - X_i)}{\sigma_i} dW_i$$

$$d\Pi = \lambda \sum_i \frac{k_i}{2} \xi_i^2 dt + \lambda \sum_i \sqrt{\frac{k_i}{2}} \xi_i dW_i \quad \xi_i = \frac{m - X_i}{\sigma_i} \sqrt{2k_i}$$

$$\langle d\Pi \rangle = \frac{\lambda N}{2} \left(\frac{\sum_i k_i}{N} \right) dt \quad ; \quad \langle (d\Pi)^2 \rangle - \langle d\Pi \rangle^2 = \frac{\lambda^2 N}{2} \left(\frac{\sum_i k_i}{N} \right) dt$$

$$\text{Annualized Sharpe Ratio} = \sqrt{\frac{N}{2} \cdot \left(\frac{\sum_i k_i}{N} \right)} = \sqrt{\frac{N \bar{k}}{2}}$$

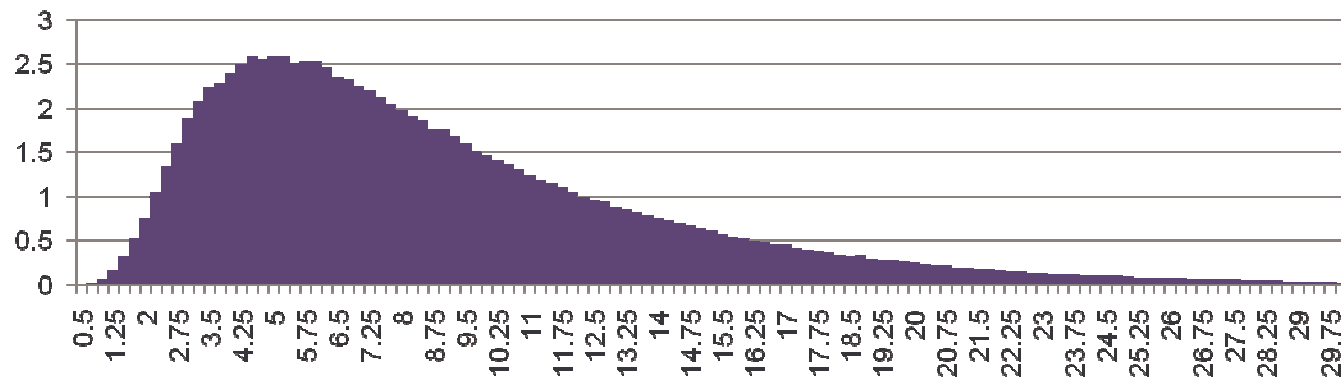
Statistics on the Estimated OU Parameters

ETF	Abs(Alpha)	Beta	Kappa	Reversion days	EquiVol	Abs(m)
HHH	0.20%	0.69	38	7	4%	3.3%
IYR	0.11%	0.90	39	6	2%	1.8%
IYT	0.18%	0.97	41	6	4%	3.0%
RKH	0.10%	0.98	39	6	2%	1.7%
RTH	0.17%	1.02	39	6	3%	2.7%
SMH	0.19%	1.01	40	6	4%	3.2%
UTH	0.09%	0.81	42	6	2%	1.4%
XLF	0.11%	0.83	42	6	2%	1.8%
XLI	0.15%	1.15	42	6	3%	2.4%
XLK	0.17%	1.03	42	6	3%	2.7%
XLP	0.12%	1.01	42	6	2%	2.0%
XLV	0.14%	1.05	38	7	3%	2.5%
XLY	0.16%	1.03	39	6	3%	2.5%
Total	0.15%	0.96	40	6	3%	2.4%

Average over 2006-2007

Mean reversion days: how long does it take to converge?

Distribution of reversion days



$$T_{\text{days}} = 252/k$$

	Days
Max	30
75 %	11.4
Median	7.5
25 %	4.9
Min	0.5
Fast days	36%

Fast days : Percentage of faster mean reversion than 7 days