

FX Exposure

1. Measuring Exposure

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FX Risk Management

- **Exposure (Risk)**

- At the firm level, currency risk is called *exposure*.

- **Three areas**

(1) *Transaction exposure*: Risk of transactions denominated in FC with a payment date or maturity.

(2) *Economic exposure*: Degree to which a firm's expected cash flows are affected by unexpected changes in S_t .

(3) *Translation exposure*: Accounting-based changes in a firm's consolidated statements that result from a change in S_t . Translation rules create accounting gains/losses due to changes in S_t .

We say a firm is “*exposed*” or *has exposure* if it faces currency risk.

Example: Exposure.

A. *Transaction exposure.*

Swiss Cruises, a Swiss firm, sells cruise packages priced in USD to a broker. Payment in 30 days.

B. *Economic exposure.*

Swiss Cruises has 50% of its revenue denominated in USD and only 20% of its cost denominated in USD. A depreciation of the USD will affect future CHF cash flows.

C. *Translation exposure.*

Swiss Cruises obtains a USD loan from a U.S. bank. This liability has to be translated into CHF following Swiss accounting rules. ¶

Q: How can FX changes affect the firm?

- *Transaction Exposure*

- Short-term CFs: Existing contract obligations.

- *Economic Exposure*

- Future CFs: Erosion of competitive position.

- *Translation Exposure*

- Revaluation of balance sheet (Book Value vs Market Value).

Measuring Transaction Exposure

- Transaction exposure (TE) is easy to identify and measure.
 - Identification: Transactions denominated in FC with a **fixed** future date
 - Measure: Translate identified FC transactions to DC using S_t .

$$TE_{j,t} = \text{Value of a fixed future transaction in FC}_j * S_t$$

Example: Swiss Cruises.

Sold cruise packages for USD 2.5 million. Payment: 30 days.

Bought fuel oil for USD 1.5 million. Payment: 30 days.

$S_t = 1.45 \text{ CHF/USD}$.

Thus, the net transaction exposure in USD 30 days is:

$$\begin{aligned} \text{Net } TE_{j=USD} &= (\text{USD } 2.5\text{M} - \text{USD } 1.5\text{M}) * 1.45 \text{ CHF/USD} \\ &= \text{USD } 1\text{M} * 1.45 \text{ CHF/USD} = \text{CHF } 1.45\text{M}. \quad \P \end{aligned}$$

• Netting

An MNC has many transactions, in different currencies, with fixed futures dates. Since TE is denominated in DC, all exposures are easy to consolidate in one single number: Net TE (NTE).

$$\text{NTE} = \text{Net } TE_t = \sum_{j=1}^J TE_{j,t} \quad j = \text{EUR, GBP, JPY, BRL, MXN, ...}$$

- NTE is reported by fixed date: up to 90 days, more than 90-days, etc.

Note: Since currencies are correlated, firms take into account **correlations** to calculate how changes in S_t affect Net TE \Rightarrow **Portfolio Approach**.

Example: A U.S. MNC: Subsidiary A with CF(in EUR) > 0
 Subsidiary B with CF(in GBP) < 0

Since $\rho_{\text{GBP,EUR}}$ is very high and positive, NTE may be very low. \P

\Rightarrow Hedging decisions are usually made based on exposure of the **portfolio**.

• Netting - Correlations

Example: Swiss Cruises.

Net Inflows (in USD): **USD 1 million**. Due: 30 days.

Loan repayment: **CAD 1.50 million**. Due: 30 days.

$S_t = 1.47 \text{ CAD/USD}$.

$\rho_{\text{CAD,USD}} = .843$ (monthly from 1971 to 2017)

Swiss Cruises considers NTE to be close to zero. ¶

Note 1: Correlations vary a lot across currencies. In general, **regional** currencies are highly correlated.

From **2000-2017**,

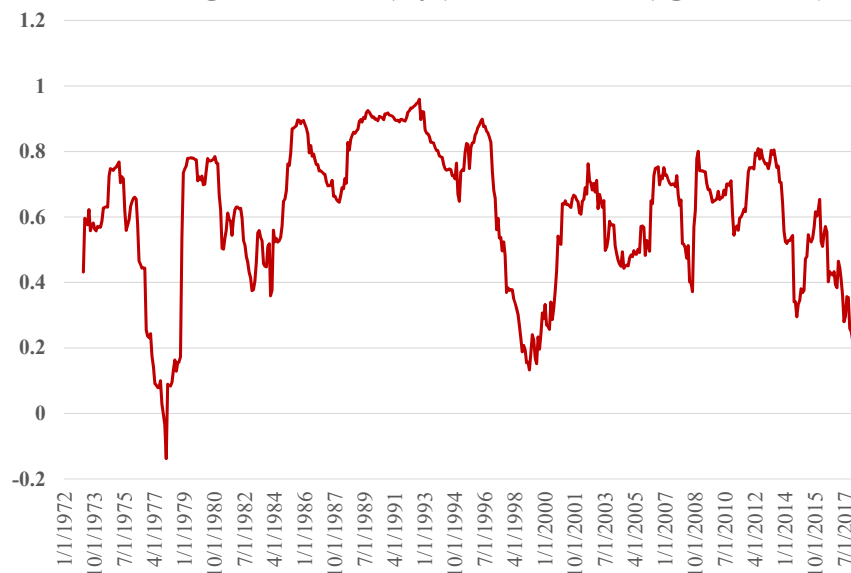
$$\rho_{\text{GBP,NOK}} = \mathbf{0.58}$$

$$\rho_{\text{GBP,JPY}} = \mathbf{0.04}$$

Note 2: Correlations also vary over time.

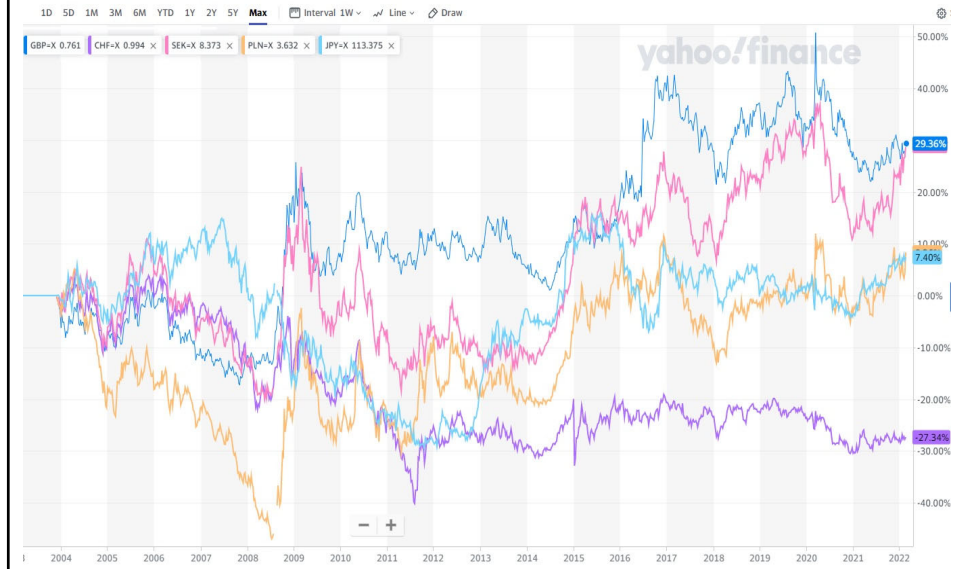
• Netting - Correlations

Rolling Correlation (2-yr) GBP & NOK (against USD)



• Netting - Correlations

On average, currencies from developed countries tend to move together...
But, not all and not always.



• Q: How does TE affect a firm in the future?

Firms are interested in how TE will change in the future, say, in T days when transaction will be settled.

- Firms do not know S_{t+T} , they need to forecast $S_{t+T} \Rightarrow E_t[S_{t+T}]$
- Once we forecast $E_t[S_{t+T}]$, we can forecast $E_t[TE_{t+T}]$:

$$E_t[TE_{t+T}] = \text{Value of a fixed future transaction in FC} * E_t[S_{t+T}]$$
- $E_t[S_{t+T}]$ has an associated standard error, which can be used to create a range (or interval) for S_{t+T} & TE.
- Risk management perspective:
 How much DC can the firm spend on account of a FC inflow in T days?
 How much DC will be needed to cover a FC outflow in T days?

Range Estimates of TE

- S_t is very difficult to forecast. Thus, a range estimate for NTE provides a useful number for risk managers.

The smaller the range, the lower the sensitivity of NTE.

- Three popular methods for estimating a range for NTE:
 - (1) Ad-hoc rule (say, $\pm 10\%$)
 - (2) Sensitivity Analysis (or simulating exchange rates)
 - (3) Assuming a statistical distribution for exchange rates.

• Ad-hoc Rule

Many firms use an *ad-hoc* (“arbitrary”) rule to get a range: $\pm X\%$ (for example, a 10% rule)

Simple and easy to understand: Get TE and add/subtract $\pm X\%$.

Example: 10% Rule.

SC has a Net TE = CHF 1.45M due in 30 days

⇒ if S_t changes by $\pm 10\%$, NTE changes by \pm CHF 145,000. ¶

Note: This example gives a range for NTE:

NTE \in [CHF 1.305M; CHF 1.595M]

Risk Management Interpretation: A risk manager will only care about the lower bound. If SC is counting on the USD 1M inflow to pay CHF expenses, these expenses should not exceed CHF 1.305M. ¶

• Sensitivity Analysis

Goal: Measure the sensitivity of TE to different exchange rates.

Example: Sensitivity of TE to extreme forecasts of S_t .

Sensitivity of TE to randomly simulate thousands of S_t .

Data: 20 years of monthly CHF/USD % changes (ED)

Mean (μ)	-0.00152	$\mu_m = -0.152\%$
Standard Error	0.00202	
Median	-0.00363	
Mode	#N/A	
Stand Deviation (σ)	0.03184	$\sigma_m = 3.184\%$
Sample Variance (σ^2)	0.00101	
Kurtosis	0.46327	
Skewness	0.42987	
Range	0.27710	
Minimum	-0.11618	
Maximum	0.15092	
Sum	0.0576765	
Count	248	

• Sensitivity Analysis – Extremes (Worst Case & Best Case)

Example: Extremes for Swiss Cruises Net TE (CHF/USD)

ED of S_t monthly changes over the past 20 years (1994-2014).

Extremes: **15.09%** (on October 2011) and **-11.62%** (on Jan 2009).

SC's net receivables in FC: **USD 1M**.

(A) *Best case scenario*: largest appreciation of USD: **0.1509**

NTE: **USD 1M** * **1.45 CHF/USD** * (1 + **0.1509**) = **CHF 1,668,805**.

(B) *Worst case scenario*: largest depreciation of USD: **-0.1162**

NTE: **USD 1M** * **1.45 CHF/USD** * (1 + (**-0.1162**)) = **CHF 1,281,510**.

That is,

$$\text{NTE} \in [\text{CHF } 1,281,510; \text{CHF } 1,668,805]$$

Note: If Swiss Cruises is counting on the USD 1M to cover CHF expenses, the expenses to cover should not exceed **CHF 1,281,510**. ¶

• Sensitivity Analysis – Simulation

Managers may consider the previous range, based on extremes, too conservative:

$$\text{NTE} \in [\text{CHF } 1,281,510; \text{CHF } 1,668,805].$$

⇒ Probability of worst case scenario is low: Only once in 240 months!

Under more likely scenarios, a firm may be able to cover more expenses.

A more realistic range can be constructed through sampling from the ED.

Example: Simulation for SC's Net TE (CHF/USD) over one month.

(i) Randomly pick 1,000 monthly $e_{f,t+30}$'s from the ED.

(ii) Calculate S_{t+30} for each $e_{f,t+30}$ selected in (i).

$$(\text{Recall: } S_{t+30} = 1.45 \text{ CHF/USD} * (1 + e_{f,t+30}))$$

(iii) Calculate **TE** for each S_{t+30} . (Recall: **TE** = USD 1M * S_{t+30})

(iv) Plot the 1,000 **TE**'s in a histogram. (Simulated TE distribution.)

Example (continuation): In excel, using Vlookup function

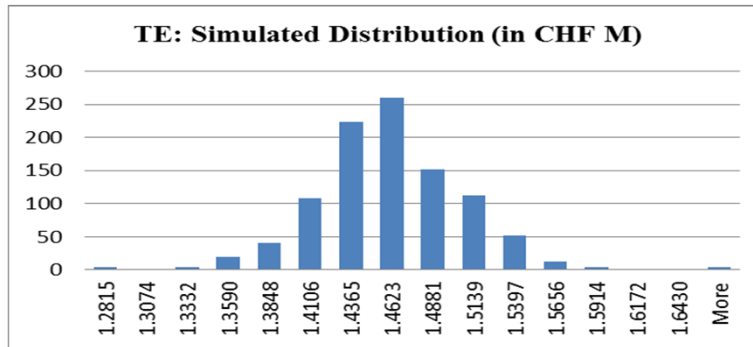
(i) Randomly draw $e_{f,t} = e_{f,\text{sim},1}$ from ED: Observation 19: $e_{f,t+30} = 0.0034$

(ii) Calculate $S_{\text{sim},1}$: $S_{t+30} = 1.45 \text{ CHF/USD} * (1 + .0034) = 1.4549$

(iii) Calculate $\text{TE}_{\text{sim},1}$: **TE** = USD 1M * $S_{t+30} = 1,454,937.57$

(iv) Repeat (i)-(iii) 1,000 times. Plot histogram. Construct a $(1-\alpha)\%$ C.I.

Lookup cell	Random Draw		Draw s_sim		TE(sim)
	$e_{f,t}$	with Randbetween	with Vlookup	S_{sim}	
1					
2	0.0025	19	0.0034	1.4549	1,454,937.57
3	-0.0027	147	-0.0104	1.4349	1,434,895.83
4	0.0001	99	0.0125	1.4682	1,468,189.96
5	-0.0443	203	-0.0584	1.3653	1,365,272.92
6	-0.0017	82	-0.0727	1.3446	1,344,597.25
7	-0.0031	4	0.0001	1.4502	1,450,168.79
8	-0.0227	67	-0.0226	1.4172	1,417,218.22
9	-0.0099	136	0.0095	1.4638	1,463,838.02
10	0.0098	232	0.0191	1.4777	1,477,749.46



Based on this simulated distribution, we can estimate a 95% range (leaving 2.5% observations to the left and 2.5% observations to the right)

$$\Rightarrow \text{NTE} \in [\text{CHF } 1.3661 \text{ M}; \text{ CHF } 1.5443 \text{ M}]$$

Practical Application: If SC expects to cover expenses with this USD inflow, the maximum amount in CHF to cover, using this 95% CI, should be **CHF 1,366,100.** ¶

• **Aside: How many draws in the simulations?**

Usually, we draw until the CIs do not change a lot.

Example: 1,000 and 10,000 draws

For the SC example, we drew **1,000** scenarios to get a 95% C.I.:

$$\Rightarrow \text{NTE} \in [\text{CHF } 1.3661 \text{ M}; \text{ CHF } 1.5443 \text{ M}]$$

Now, we draw **10,000** scenarios and determined the following 95% C.I.:

$$\Rightarrow \text{NTE} \in [\text{CHF } 1.3670 \text{ M}; \text{ CHF } 1.5446 \text{ M}]$$

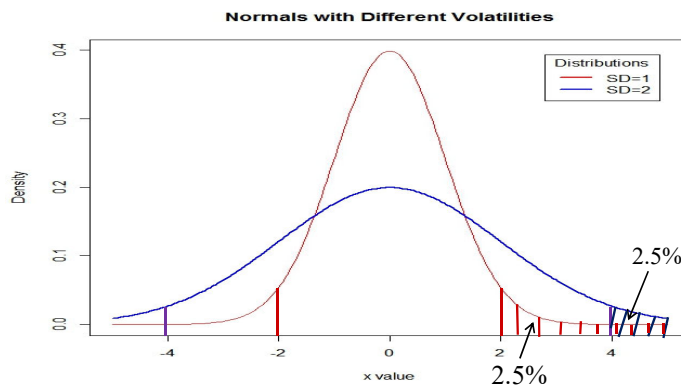
- Not a significant change in the range: 1,000 simulations seem enough.

• Assuming a Distribution

CI's based on an assumed distribution provide a range for TE.

For example, a firm assumes that $e_{f,t} \sim N(\mu, \sigma^2)$. (“~” = follows)
 \Rightarrow construct a $(1 - \alpha)\%$ CI: $[\mu \pm |z_{\alpha/2}| \sigma]$. ($|z_{\alpha/2}| = z_{1-\alpha/2}$)

Usual α 's: $\alpha = .05 \Rightarrow |z_{.025}| = 1.96 (\approx 2)$
 $\alpha = .02 \Rightarrow |z_{.01}| = 2.33$



• Assuming a Distribution – Normal for $e_{f,t}$

Example: CI range based on a Normal distribution.

Swiss Cruises believes that CHF/USD monthly changes follow a normal distribution. SC estimates:

μ = Monthly mean = **-0.00152** \approx -0.15%

σ^2 = Monthly variance = 0.001014 ($\Rightarrow \sigma =$ **0.03184**, or 3.18%)

$e_{f,t} \sim N(\text{b}-0.00152, \text{b}0.03184^2)$ $e_{f,t}$ = CHF/USD monthly changes.

SC builds a 95% CI for CHF/USD monthly changes:

$$[-0.00152 \pm 1.96 * 0.03184] = [\text{b}-0.06393; 0.06089].$$

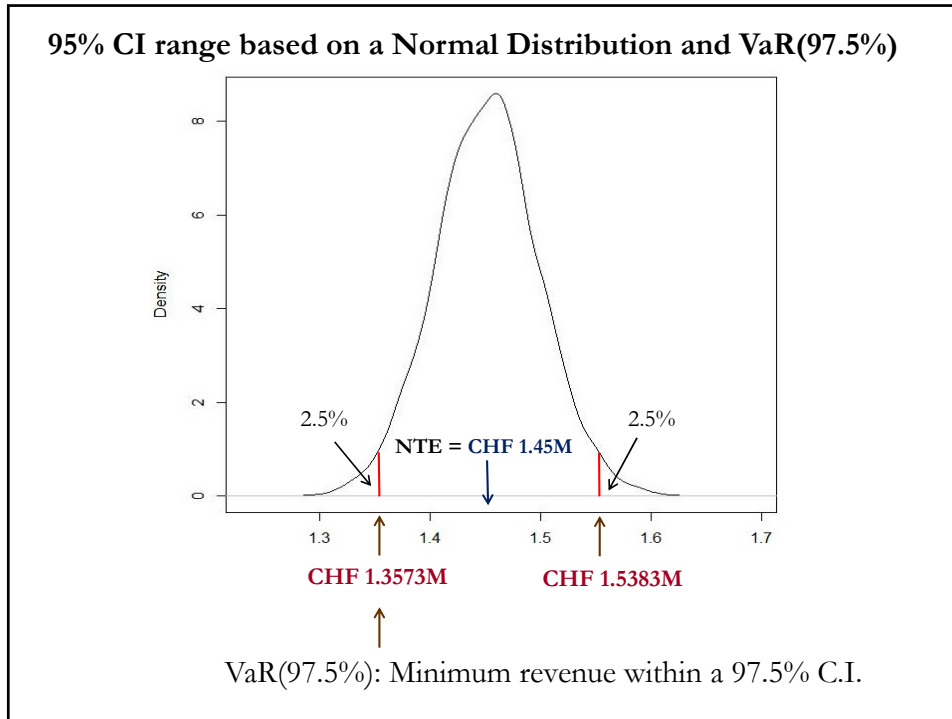
Based on this range for $e_{f,t}$, we derive bounds for the net TE:

(A) Upper bound

NTE: **USD 1M** * **1.45 CHF/USD** * $(1 + 0.06089) =$ **CHF 1,538,291**.

(B) Lower bound

NTE: **USD 1M** * **1.45 CHF/USD** * $(1 + (\text{b}-0.06393)) =$ **CHF 1,357,302**.



$$\Rightarrow \text{NTE} \in [\text{CHF } 1.357 \text{ M}; \text{CHF } 1.538 \text{ M}]$$

- The lower bound, for a receivable, represents the worst case scenario within the confidence interval.

There is a **Value-at-Risk (VaR)** interpretation:

VaR: *Maximum expected loss* in a given time interval within a (one-sided) CI.

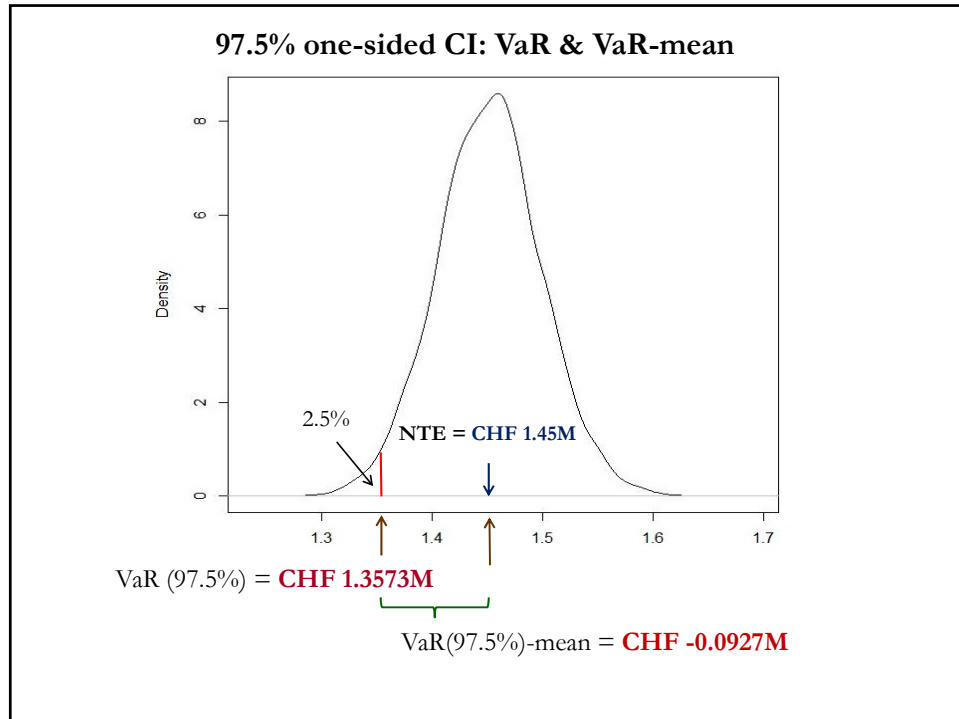
In our case, we can express the “*expected loss*” relative to today’s value:

$$\text{VaR-mean} = \text{VaR} - \text{NTE}$$

Example (continuation): The minimum revenue to be received by SC in the next 30 days, within a 97.5% CI ($\Rightarrow |z_{.025}| = 1.96$):

$$\begin{aligned} \text{VaR}(97.5\%) &= \text{CHF } 1.45\text{M} [1 + (-0.00152 - 1.96 * 0.03184)] \\ &= \text{CHF } 1,357,302. \end{aligned}$$

$$\Rightarrow \text{VaR-mean } (.975) = \text{CHF } 1.3573\text{M} - \text{CHF } 1.45\text{M} = \text{CHF } -0.0927\text{M}$$



Example (continuation): $\Rightarrow \text{NTE} \in [\text{CHF } 1.357 \text{ M}; \text{CHF } 1.538 \text{ M}]$

$\text{VaR}(97.5\%) = \text{CHF } 1,357,302$

If SC expects to cover expenses with this USD inflow, the maximum amount in CHF to cover, within a 97.5% CI, should be **CHF 1,357,302**.

$\text{VaR-mean}(97.5\%) = \text{CHF } -0.0927\text{M}$

Relative to today's valuation (or *expected valuation*, according to RWM), the maximum *expected loss* with a 97.5% "chance" is **CHF -0.0927M**. ¶

Note: We could have used a different significance level to calculate the VaR, for example 99% ($\Rightarrow |z_{.01}| = 2.33$). Then,

$$\begin{aligned} \text{VaR}(99\%) &= \text{CHF } 1.45\text{M} [1 + (-0.00152 - 2.33 * 0.03184)] \\ &= \text{CHF } 1.34023. \quad (\text{A more conservative bound.}) \end{aligned}$$

$$\Rightarrow \text{VaR-mean}(.99) = \text{CHF } 1.34023\text{M} - \text{CHF } 1.45\text{M} = \text{CHF } -0.1098\text{M}$$

• **Summary NTE for Swiss Cruises:**

- NTE = CHF 1.45M

• **NTE Range:**

◊ **Ad-hoc:**

$$\text{NTE} \in [\text{CHF } 1.305\text{M}; \text{ CHF } 1.595 \text{ M}].$$

◊ **Sensitivity Analysis:**

- Extremes: $\text{NTE} \in [\text{CHF } 1.281 \text{ M}; \text{ CHF } 1.6688 \text{ M}]$
- Simulation: $\text{NTE} \in [\text{CHF } 1.3661 \text{ M}; \text{ CHF } 1.5443 \text{ M}]$

◊ **Statistical Distribution (normal):**

$$\text{NTE} \in [\text{CHF } 1.357 \text{ M}; \text{ CHF } 1.538 \text{ M}]$$

♦ **Approximating Returns**

In general, we use *arithmetic returns*: $e_{f,t} = S_t/S_{t-1} - 1$. To change the frequency, compounding is needed.

But, if we use *logarithmic returns* –i.e., $e_{f,t} = \log(S_t) - \log(S_{t-1})$ –, changing the frequency of mean returns (μ) and return variances (σ^2) is simpler.

Let μ_b & σ_b^2 be measured in a given base frequency, say, b . Then,

$$\begin{aligned} \mu_f &= \mu_b * T, \\ \sigma_f^2 &= \sigma_b^2 * T \quad \Rightarrow \sigma_f = \sigma_b * \text{sqrt}(T) \end{aligned}$$

T = # periods of base frequency b in new frequency, f .

♦ **Approximating Returns – From monthly to daily & annual**

Example: Using monthly data, compute daily and annual mean & SD.

From previous Table (base frequency: b = monthly, arithmetic computed):

$$\mu_m = -0.00152$$

$$\sigma_m = 0.03184$$

(1) Daily (i.e., $f = d = \text{daily}$ & $T = 1/30$)

$$\mu_d = (-0.00152) * (1/30) = .0000507 \quad (0.006\%)$$

$$\sigma_d = (0.03184) * (1/30)^{1/2} = .00602 \quad (0.60\%)$$

(2) Annual (i.e., $f = a = \text{annual}$ & $T = 12$)

$$\mu_a = (-0.00152) * (12) = -0.01824 \quad (-1.82\%)$$

$$\sigma_a = (0.03184) * (12)^{1/2} = 0.110297 \quad (11.03\%)$$

Check: The annual compounded arithmetic return:

$$(1 - 0.00152)^{12} - 1 = -0.01809.$$

When arithmetic returns are low, these approximations work well. ¶

♦ **Approximating Returns – From monthly VaR to annualized VaR**

Example: Using the annualized approximation, we can also approximate an annualized VaR(97.5%) for Swiss Cruises:

$$\begin{aligned} \text{VaR}(97.5\%) &= \text{USD } 1\text{M} * 1.45 \text{ CHF/USD} * [1 + (-0.01824 - 1.96 * 0.1103)] \\ &= \text{CHF } 1,101,374. \quad \P \end{aligned}$$

Note II: Using logarithmic returns rules, we can approximate USD/CHF monthly changes by changing the sign of the CHF/USD, while the variance remains the same.

Then,

- Annualized USD/CHF mean percentage change $\approx 1.82\%$,
- Annualized USD/CHF volatility $\approx 11.03\%$

● Sensitivity Analysis – Portfolio Approach

A simulation: Draw different scenarios, pay attention to *correlations*!

Example: IBM has the following CFs in the next 90 days

	Outflows	Inflows	S_t	Net Inflows
GBP	100,000	25,000	1.60 USD/GBP	(75,000)
EUR	80,000	200,000	1.05 USD/EUR	120,000

$$\begin{aligned} \text{NTE}_0 &= \text{EUR } 120\text{K} * 1.05 \text{ USD/EUR} + (\text{GBP } 75\text{K}) * 1.60 \text{ USD/GBP} \\ &= \text{USD } 6,000 \text{ (this is our baseline case)} \end{aligned}$$

Situation 1: Assume $\rho_{\text{GBP,EUR}} = 1$. (EUR and GBP correlation is high.)

Scenario (i): EUR appreciates by 10% against the USD

$$\begin{aligned} \text{Since } \rho_{\text{GBP,EUR}} = 1, \quad S_t &= 1.05 \text{ USD/EUR} * (1 + .10) = 1.155 \text{ USD/EUR} \\ S_t &= 1.60 \text{ USD/GBP} * (1 + .10) = 1.76 \text{ USD/GBP} \end{aligned}$$

$$\begin{aligned} \text{NTE} &= \text{EUR } 120\text{K} * 1.155 \text{ USD/EUR} + (\text{GBP } 75\text{K}) * 1.76 \text{ USD/GBP} \\ &= \text{USD } 6,600. \text{ (+10\% change = USD -600)} \end{aligned}$$

● Sensitivity Analysis – Portfolio Approach

Example (continuation): with $\rho_{\text{GBP,EUR}} = 1$.

Scenario (ii): EUR depreciates by 10% against the USD

$$\begin{aligned} \text{Since } \rho_{\text{GBP,EUR}} = 1, \quad S_t &= 1.05 \text{ USD/EUR} * (1 - .10) = 0.945 \text{ USD/EUR} \\ S_t &= 1.60 \text{ USD/GBP} * (1 - .10) = 1.44 \text{ USD/GBP} \end{aligned}$$

$$\begin{aligned} \text{NTE} &= \text{EUR } 120\text{K} * 0.945 \text{ USD/EUR} + (\text{GBP } 75\text{K}) * 1.44 \text{ USD/GBP} \\ &= \text{USD } 5,400. \text{ (-10\% change = USD -600)} \end{aligned}$$

Now, we can specify a range for NTE

$$\Rightarrow \text{NTE} \in [\text{USD } 5,400; \text{USD } 6,600]$$

Note: The NTE change is exactly the same as the change in S_t . Then,

$$\text{if } \text{NTE}_0 \approx 0 \quad \Rightarrow e_{f,t} \text{ has very small effect on NTE.}$$

That is, if a firm has matching inflows and outflows in highly positively correlated currencies, then changes in S_t do not affect NTE. From a risk management perspective, this is very good.

● Sensitivity Analysis – Portfolio Approach

Example (continuation):

Situation 2: Suppose the $\rho_{\text{GBP, EUR}} = -1$ (NOT a realistic assumption!)

Scenario (i): EUR **appreciates** by 10% against the USD

Since $\rho_{\text{GBP, EUR}} = -1$, $S_t = 1.05 \text{ USD/EUR} * (1 + .10) = 1.155 \text{ USD/EUR}$
 $S_t = 1.60 \text{ USD/GBP} * (1 - .10) = 1.44 \text{ USD/GBP}$

NTE = EUR 120K * 1.155 USD/EUR + (GBP 75K) * 1.44 USD/GBP
 = USD 30,600. (410% change = USD 24,600)

Scenario (ii): EUR **depreciates** by 10% against the USD

Since $\rho_{\text{GBP, EUR}} = -1$, $S_t = 1.05 \text{ USD/EUR} * (1 - .10) = 0.945 \text{ USD/EUR}$
 $S_t = 1.60 \text{ USD/GBP} * (1 + .10) = 1.76 \text{ USD/GBP}$

NTE = EUR 120K * 0.945 USD/EUR + (GBP 75K) * 1.76 USD/GBP
 = (USD 18,600). (-410% change = USD -24,600)

Now, we can specify a range for NTE

$$\Rightarrow \text{NTE} \in [(\text{USD } 18,600); \text{USD } 30,600]$$

● Sensitivity Analysis – Portfolio Approach

Example (continuation):

Note: The NTE has ballooned. A 10% change in S_t a dramatic increase in the NTE range.

\Rightarrow Having non-matching exposures in different currencies with negative correlation is very dangerous.

Remarks:

- IBM can assume a correlation (estimated from the data). Then, draw many scenarios from a *bivariate normal distribution* to generate a simulated distribution for the NTE.

- Alternatively, IBM can just draw joint pairs from the ED. From this ED, IBM will get a range –and a VaR– for the NTE. ¶