# Value-at-Risk for Currencies (FX Spot)

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October 5, 2021

# PHP as Quote Currency

- For simplicity, we will assume that the quote currency for all exchange rates is Philippine pesos (PHP).
- Thus, when we say we have a long position on a currency CCY, we are implying that we have a long position on the currency pair CCYPHP.
- On the other hand, taking a short position on CCY means shorting the currency pair CCYPHP.
- Therefore, all historical exchange rates to be used in calculating VaR must be expressed in CCYPHP.

# Calculating $\Delta P$

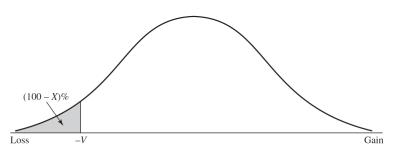
Suppose a FI holds a portfolio consisting solely of currency  $C_1$ . Let  $F_1$  be the FI's position on  $C_1$ ,  $y_{1,0}$  be the exchange rate to peso of  $C_1$  today, and  $y_1$  be the exchange rate to peso of  $C_1$  tomorrow. The peso equivalents of the position on  $C_1$  today and tomorrow are  $F_1y_{1,0}$  and  $F_1y_1$ , respectively. Hence, the change in the position's peso equivalent today and tomorrow is

$$\Delta P = F_1 y_1 - F_1 y_{1,0}$$
$$= F_1 (y_1 - y_{1,0})$$

## Dealing with Long and Short Positions

- When the bank takes a <u>long position</u> on  $C_1$ , then  $F_1$  is given a positive sign. Thus,  $\Delta P$  is positive when  $y_1 > y_{1,0}$  and negative when  $y_1 < y_{1,0}$ . This is consistent with the fact that a long position on a currency gains value when the currency's exchange rate to peso increases but loses value when the exchange rate decreases.
- On the other hand, when the bank takes a <u>short position</u> on  $C_1$ ,  $F_1$  is assigned a negative sign, Thus,  $\Delta P$  is positive when when  $y_1 < y_{1,0}$  and negative when  $y_1 > y_{1,0}$ , consistent with the fact that a short position gains value when the exchange rate decreases and loses value when the exchange rate increases.
- Making  $F_1$  a signed quantity (+ if long, if short) ensures the consistency of our interpretation of  $\Delta P$ :
  - $\Delta P > 0$  implies a gain in the value of the portfolio
  - ullet  $\Delta P < 0$  implies a loss in the value of the portfolio

#### An Illustration



**Figure 12.1** Calculation of VaR from the Probability Distribution of the Gain in the Portfolio Value

Losses are negative gains; confidence level is X%; VaR level is V.

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# Delta-Normal Approach

# Single Currency $C_1$

#### Recall: Stocks

The one-day 99% VaR, |V|, of a portfolio consisting of a long position on N on a stock with current share price  $S_0$  is

$$|V| = NS_0 \sigma \Phi^{-1}(0.99)$$

where  $\sigma$  is the daily volatility of stock returns.

#### Formula for VaR

The one-day 99% VaR, |V|, of a portfolio consisting of a position  $F_1$  on currency  $C_1$  is

$$|V| = |F_1|y_{1,0}\sigma_1\Phi^{-1}(0.99)$$

where  $y_{1,0}$  is the current exchange rate of  $C_1$  to PHP and  $\sigma_1$  is the daily volatility of exchange-rate returns.<sup>1</sup>

 $^{1}$ The formula can be modified for any  $100\alpha\%~{
m VaR}$ .

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For an n-currency portfolio, we denote the currencies as  $C_1, C_2, \ldots, C_n$  each with the current exchange rate to PHP denoted by  $y_{1,0}, y_{2,0}, \ldots, y_{n,0}$  respectively.

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#### Undiversified VaR

To compute for the undiversified  $99\%\,VaR$ , calculate the VaR for each currency in the portfolio and take the sum

$$|V_u| = \sum_{i=1}^n |F_i| y_{i,0} \sigma_i \Phi^{-1}(0.99).$$

#### Diversified VaR

Let  $a_i = F_i y_{i,0}, i = 1, 2, \dots, n$ 

 $\boldsymbol{a}$  be the vector of  $a_i$ 's

 $\Sigma$  be the covariance matrix

The diversified  $99\% \text{ VaR } |V_d|$  is given by

$$|V_d| = \sigma_P \Phi^{-1}(0.99)$$

where

$$\sigma_P^2 = \boldsymbol{a}^T \boldsymbol{\Sigma} \boldsymbol{a}.$$

## N-day VaR and EWMA model

The approaches for calculating N-day VaR and the procedure for computing VaR using Exponentially-Weighted Moving Average (EWMA) model also easily translate to a portfolio of currencies. The returns on the exchange rates are used instead of the returns on stock prices.

#### Historical Simulation

Suppose the historical data consists of the most recent M+1 days of historical exchange rates to PHP for currencies in the portfolio.

#### Procedure using returns on individual currencies

• Generate possible 1-day returns based on the previous 1-day returns.

 $R_{1,i,j} = \ln\left(\frac{y_{i,j}}{y_{i,j-1}}\right)$ 

2 Compute for the possible change in portfolio value using the generated 1-day return.

$$\Delta P^{(j)} = \sum_{i=1}^{n} F_i y_{i,0} R_{1,i,j}$$

3 The 1-day 99% VaR is the  $(0.01M)^{\text{th}}$  smallest value.

# BRW Approach

# BRW Approach: Single Currency Case

#### Step 1: Assignment of Weights

- ① Calculate the observed one-day continuously compounded returns  $R_{1,0}, R_{1,1}, \ldots, R_{1,M-1}$  from the exchange rates of currency  $C_1$  or a portfolio of currencies, where  $R_{1,j} = \ln(y_{1,j}/y_{1,j-1})$  is the rate of return observed j days ago.
- ② Compute for scenarios of changes in portfolio value  $\Delta P^{(j)} = F_1 y_{1,0} R_{1,j}$  for  $j=0,1,\ldots,M-1$ .
- 3 Let  $0<\lambda_1<1$  be constant and let  $w_{1,j}$  be the weight assigned to  $R_{1,j}$  for all  $j=0,1,\ldots,M-1$ . The weights are assumed to decay exponentially, so we assume that  $w_{1,j+1}=\lambda_1w_{1,j}$ . Making sure that the weights add up to 1, it can be shown that

$$w_{1,j} = \left(\frac{1-\lambda_1}{1-\lambda_1^M}\right) \lambda_1^j, \qquad j = 0, 1, \dots, M-1$$

## BRW Approach

#### Step 2: Constructing the Empirical Distribution

After assigning the weights to each element of the scenario space for  $\Delta P$ , we then arrange the  $\Delta P^{(j)}$ 's and  $w_j$ 's in increasing order of  $\Delta P^{(j)}$ . We denote the resulting sequence as follows:

The empirical cumulative distribution at  $\Delta \widetilde{P}^{(j)}$ , which denoted by  $\psi_j$ , is determined by the accumulation of weights

$$\psi_j = \sum_{k=0}^{j} w_{(k)}, \qquad j = 0, 1, \dots, M - 1$$

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# BRW Approach

#### Step 3: Calculating the 1-day 100lpha% VaR

The  $100\alpha\%$  value-at-risk is obtained by first identifying successive ordered pairs  $(\Delta \tilde{P}^{(k)}, \psi_k)$  and  $(\Delta \tilde{P}^{(k+1)}, \psi_{k+1})$  such that

$$\psi_k < 1 - \alpha < \psi_{k+1}. \tag{1}$$

Let V be the linearly interpolated value of  $\Delta P$  from the ordered pairs  $(\Delta \tilde{P}^{(k)},\psi_k)$  and  $(\Delta \tilde{P}^{(k+1)},\psi_{k+1})$  corresponding to the value  $\psi=1-\alpha.$  The  $100\alpha\%$  value-at-risk is therefore given by |V|.

# BRW Approach: **n**-asset Portfolio

#### Procedure using returns on individual currencies

- From the historical exchange rates, we can construct M scenarios for the one-day continuously compounded return  $R_{1,i}$  on the ith exchange rate.
- ② Thus, M scenarios can also be constructed for  $\Delta P_i$

$$\Delta P_{i,j} = F_i y_{i,0} R_{1,i,j}.$$

**1** The scenario space for  $\Delta P$  is given by

$$\Delta P^{(j)} = \sum_{i=1}^{n} \Delta P_{i,j} = \sum_{i=1}^{n} F_i y_{i,0} R_{1,i,j}$$

We can then resume the BRW Approach at Step 2 using  $\{\Delta P^{(j)}\}$  as the  $\{\Delta P\}$  scenarios.

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