

## Scour depth evaluation of a bridge with a complex pier foundation

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### **Abstract**

A scour depth prediction formula for a river bridge is established using experimental data in which the effects of the pier, pile-cap and pile group are considered. More than 170 experimental data entries, including different pier structural sizes, flow depths and soil covering depths, are collected and verified by existing formulae, which failed to deliver a promising prediction. A machine learning prediction model was then developed to enhance the accuracy. For application purpose, a sequential quadratic programming optimization was adopted to construct an explicit prediction formula. The MAPE was significantly improved from 102.8 to 28.9. The results indicate that the proposed formula can simultaneously satisfy the requirements of accuracy and simplicity. The proposed formula has the advantages of being conceptually consistent with observed scour behaviors and provides a solid scour depth prediction, which is an important and critical step in the bridge safety evaluation if floods are considered.

Keywords: scour, flood, complicated foundation, optimization, machine learning

### **1.0 Introduction**

The rivers in Taiwan mostly start at the mountain areas at an altitude above 3000 m and are often shorter than 200 m. The rivers and streams are often steep with rapid currents. In addition, due to the uneven rainfall distribution over space and time, the rainfall at wet and dry periods varies significantly such that rivers are likely to cause

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floods during the wet periods, leading to river bed scouring and debris flow, etc. Bridges are important infrastructures for traffic connecting two shores of rivers. However, bridge structures built in rivers often block the river courses, changing the flow conditions and leading to local scouring of the piers. Consequently, the river bridge foundations are exposed and damaged, endangering the lives and safety of road users. According to Andrić and Lu (2016), the primary reason for bridge damage in the U.S. is related to flooding. According to a report of Construction Research Institute in Taiwan, bridges in Taiwan also have the same trend. Therefore, bridge safety evaluation against floods has attracted substantial attention of many scholars and engineers. River bed scouring can mainly be divided into the following three types: general scour, contraction scour and local scour. The content of this paper focuses on local scour only, which is generally considered to be the most important part for bridge safety. Most of the pier scour research has focused on the scour with uniform piers (Salim and Jones, 1996), which had not considered the impacts of the pile-caps and pile groups on the scour depth. A non-uniform pier is one for which the cross-sectional dimension varies over the length of the pier. In the early engineering practice in Taiwan, the scour formula with a uniform foundation was often used. In reality, most of the bridges lack uniform piers such that the applicability of the uniform pier formulae is inadequate. Thus, this study focuses on building an accurate non-uniform (complicated foundation) scour formula. For scours of non-uniform piers, many important impact factors should be considered, such as the soil covering depth, pier width, pile-cap width, flow velocity, riverbed materials, scouring period and so on. The factors affecting the pier scour can be categorized as the pier geometry, flow property, material characteristic at the riverbed and scour lag, etc., which are described in detail below:

### **1.1 Pier geometry**

The influencing factors include the pier width perpendicular to the flow direction ( $b_c$ ), flow attack angle ( $\theta$ ), pile-cap width ( $b_{pc}$ ), and soil covering height (level of the top surface of the pile cap below the surrounding bed level,  $Y$ ). When the pier width ( $b_c$ ) increases, the scour depth ( $d_s$ ) also increases. If the piers are aligned with flow, the pier length ( $L$ ) has no obvious impact on the scour depth ( $d_s$ ). If a uniformly circular pier is considered, the flow attack angle ( $\theta$ ) and pier length ( $L$ ) have no influence on the scour depth ( $d_s$ ). Imamoto & Ohtoshi (1987) used the scour hole geometric similarity characteristic method while considering the horseshoe vortex and sediment transport to simulate scouring of non-uniform piers. They found that when the non-uniform ratio is greater (such as the difference between the pier width,  $b_c$  and pile-cap width,  $b_{pc}$ ), the pier scour depth is smaller. Melville & Raudkivi (1996) divided non-uniform piers into the following 3 configurations based on the soil covering depth ( $Y$ ): (1) pile-cap is below the bottom of the scour hole (Zone 1,  $Y/b_c > 2.4$ ), (2) pile-cap top is within the scour hole (Zone 2,  $2.4 \geq Y/b_c \geq 0$ ), and (3) pile-cap top is above the bed level (Zone 3,  $Y/b_c < 0$ ), as shown in Fig. 1. Compared to the scour results of a uniform pier, their experimental results indicated that Zone 1 does not affect the scour, Zone 2 reduces the scour and Zone 3 increases the scour depth.

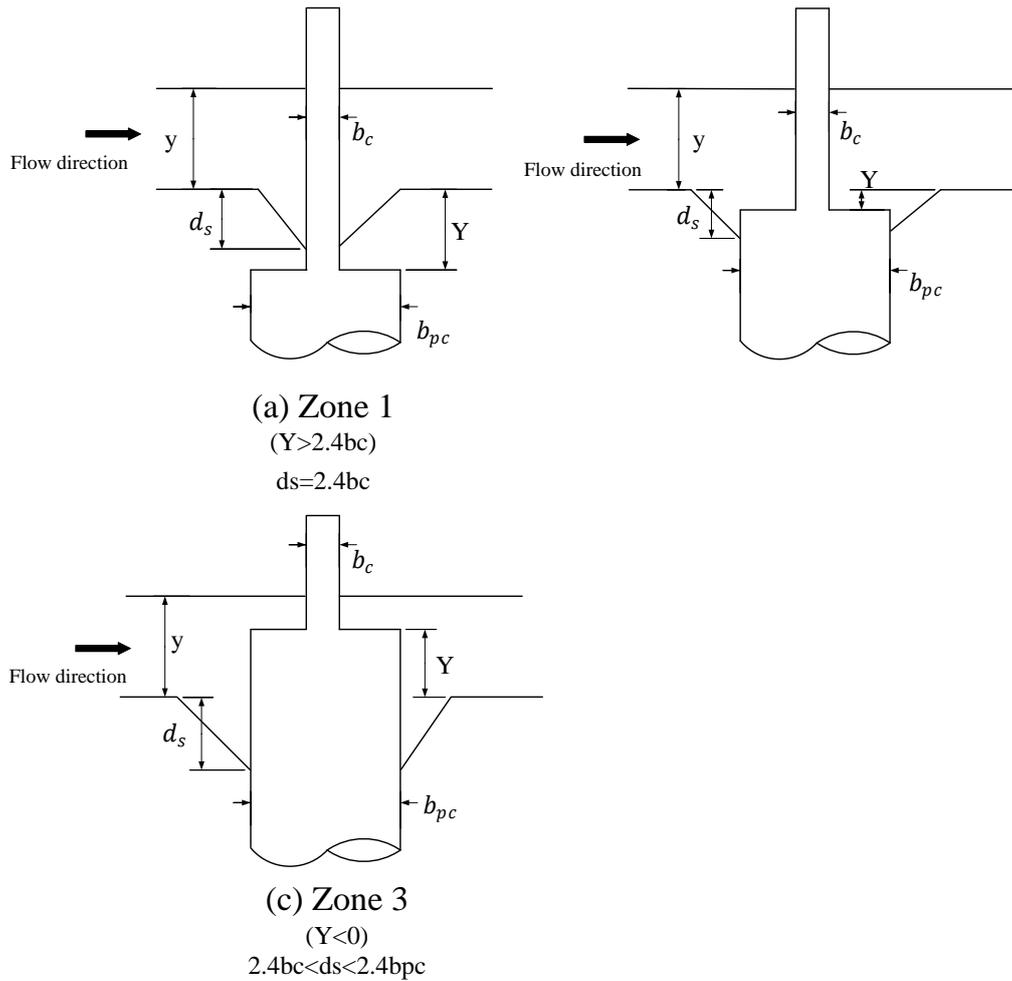


Fig. 1 Configurations of non-uniform circular piers (Melville & Raudkivi, 1996)

**1.2 Flow property**

The influential factors include the fluid density ( $\rho$ ), flow velocity ( $V$ ), flow depth ( $y$ ), and gravitation acceleration ( $g$ ). Depending on the magnitude of velocity ( $V$ ), the scour can be divided into two types, clear-water and live-bed scour (Raudkivi, 1986). In the case of clear-water scouring, the scour depth increases as a function of the flow velocity without sediment movement. In the case of live-bed scouring, because the flow velocity exceeds the critical velocity of sediment movement ( $V_c$ ), sediment transports across the bed surface, complicating the scour status (Wang et al. 2016). As the velocity exceeds the threshold velocity ( $V_c$ ), the scour depth first decreases and then increases to a second peak (Melville, 2008). As a result, average scour depth of

the live-bed scour is smaller than that of the clear-water scour depth (Melville & Coleman, 2000). Because of this, the clear-water scour depth is often adopted as the primary factor for bridge safety evaluation. In this study, in addition to considering the existing data on the clear-water scour, 4 clear-water scours are conducted in the Hydrotech Research Institute of the National Taiwan University, and all of the data are used to build the proposed formula.

The flow depth ( $y$ ) is typically standardized based on the pier width ( $b_c$ ), meaning that the value of ( $y/b_c$ ) is used to measure its impact on the scour depth. When the value of  $y/b_c$  is greater, the impact on the scour depth is greater and vice versa. Raudkivi & Ettema (1983) reported that if the value of  $y/b_c$  is greater than 3~4, the impact of the change of flow depth on the scour depth can be ignored (i.e., deep-water). The Reynolds number is often considered as one of the factors impacting the scour depth, as shown in Eq. (1).

$$d_{se} = 0.00073 \times R^{0.619} \quad (1)$$

where  $d_{se}$  refers to the equilibrium scour depth and  $R$  refers to the Reynolds number. According to Eq. (1), the scour depth increases along with increases in the Reynolds number. However, once it is increased to a particular value, the scour depth drops and the maximum value obtained at that particular value is the equilibrium scour depth. Compared to the Reynolds number, the Froude number, a dimensionless parameter representing the relative importance of the inertia and gravity effects, is often considered as a more important factor affecting the scour depth (Jain and Fisher, 1980). Hydraulic Engineering Circular No. 18 (HEC-18, 2012) is another example in which the Froude number, instead of Reynolds number, is incorporated into the prediction formula, as shown in Eq. (2)

$$\frac{d_s}{y} = \left[ 2.0K_1K_2K_3 \left( \frac{b_c}{y} \right)^{0.65} \left( \frac{V}{\sqrt{gy}} \right)^{0.43} \right] \quad (2)$$

wherein  $d_s$  is the scour depth;  $K_1$  refers to the pier shape correction factor;  $K_2$  refers to the correction coefficient for the angle of attack of flow;  $K_3$  refers to the river bed material correction coefficient;  $b_c$  refers to the pier width perpendicular to the flow; and  $\frac{V}{\sqrt{gy}}$  refers to the Froude number (Fr).

### **1.3 River bed material characteristics**

The influencing factors include the median grain size ( $d_{50}$ ), river bed material standard deviation ( $\sigma_g$ ), river bed material density ( $\rho_s$ ), critical velocity of sediment movement ( $V_c$ ) and so on. When the river bed material grain size is greater, the scour resistance is increased, resulting in a smaller local scour depth and vice versa. For example, Raudkivi & Ettema (1977) showed that when  $b_c/d_{50} > 50$ , classified as the fine grain river bed, the scour depth decreases along with decreasing  $b_c/d_{50}$ . In addition to the size of the bed material, its roughness also affects the local scour depth through the critical velocity of sediment movement ( $V_c$ ). When the river bed material grain size distribution is uneven, the armoring phenomena at the surface of the river bed material surface is formulated such that critical velocity of sediment movement is increased, decreasing the scour depth. Raudkivi & Ettema (1977) reported that during the clear-water scour, the local scour depth is significantly reduced along with an increase in the  $\sigma_g$ . When  $\sigma_g$  is greater than 1.3, the armoring phenomena are initiated.

### **1.4 Scour lag**

The equilibrium of the scour depth ( $d_{se}$ ) was mainly affected by the effect of the fluid flow and sediment transportation. The scour depth could reach 50~80% of the equilibrium scour depth by approximately 10% of the equilibrium scour time.

Melville & Chiew (1999) suggested that when the change in the scour depth within 24 hours does not reach 5% of the pier width ( $b_c$ ), the scour depth is the equilibrium scour depth ( $d_{se}$ ), and its corresponding time refers to the equilibrium scour time ( $t_e$ ). As shown in Fig. 2, compared to the live-bed scour, the time for the clear-water scour to reach equilibrium is slower. However, the live-bed scour would generate irregular vibrations due to the continuous transport of the river bed sediments. The equilibrium scour time ( $t_e$ ) is also affected by the flow velocity, as shown in Fig. 3.

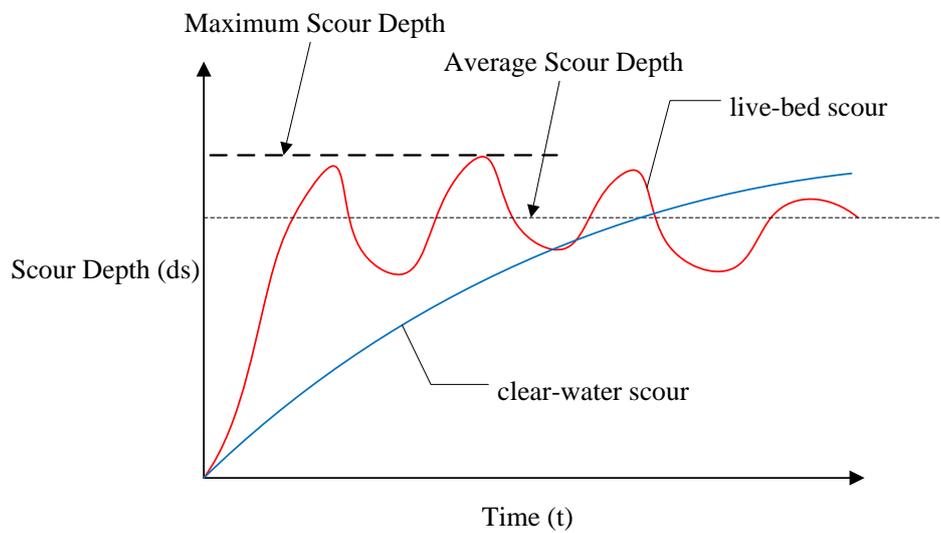


Fig. 2 Schematic view of the relationship between the scour depth and time (Raudkivi 1986)

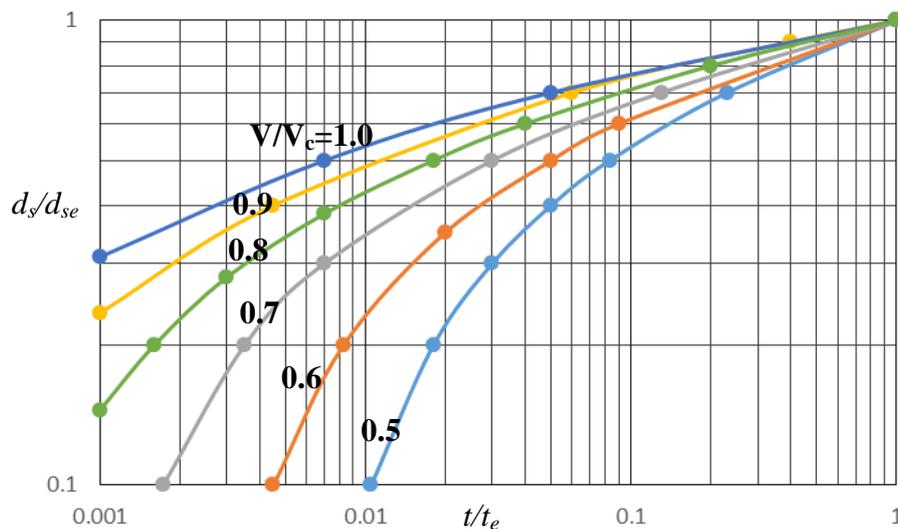


Fig. 3 Relationship between the scour depth ( $d_s$ ), equilibrium scour time ( $t_e$ ) and flow

velocity ( $V/V_c$ )

Many scholars have proposed approaches to calculate the scour depth for a pier with a non-uniform width. Among these, HEC-18 (2012) and Melville & Coleman (2000) comprehensively include the aforementioned scour factors and are more popular. As a result, they are selected as the baseline calculation in this study. HEC-18 divides the non-uniform piers into three parts (pier, pile-cap and pile group) and uses the linear superposition to predict the scour depth. Melville & Coleman (2000) take advantage of the existing formula of a uniform pier such that it would be required to obtain the equivalent pier width ( $b_e$ ) for a non-uniform pier prior to further calculation. This study first collects relevant experimental data, including data documented in the literature and new experiments. Sensitivity analysis is conducted on the experimental data to determine the important impact factors, which is followed by use of the least-square support vector machine (LS-SVM) to calculate the scour depth. Compared to the previous calculation methods (HEC-18 and Melville & Coleman, 2000), LS-SVM significantly increases the prediction accuracy. However, engineers are relatively more familiar with the utilization of formulae during the design compared to artificial intelligence (such as LS-SVM). As a result, the formula proposed by Melville & Coleman (2000) is further used with the optimization method to find the relative weight of each impact factor and enhance the prediction accuracy. The result indicates that the accuracy significantly increases, although it is slightly less than that of LS-SVM. The following provides descriptions on the existing scour depth calculation method, the artificial intelligence method (LS-SVM) adopted here, the calculation procedure of the proposed formula and the results of the analysis on the collected data.

## **2.0 Material and Methods**

Although the current study develops a method to predict the scour depth based on the following existing methods, please note other approaches are available. For example, Najafzadeh and Azamathulla (2013) proposed a quadratic polynomial of group method of data handling (GMDH) network, improved by the back propagation algorithm, to predict scour depth around bridge piers. Najafzadeh et al. (2016) integrated gene-expression programming (GEP), evolutionary polynomial regression (EPR), and model tree (MT) to predict the scour depth around bridge piers with debris effects.

### **2.1 Existing method - the HEC-18 approach**

Eq. (2) refers to the scour formula proposed by HEC-18 for a uniform pier. For the foundation with a non-uniform width, HEC-18 divides the pier structure into three parts, as shown in Fig. 4, and these three parts are the pier ( $y_{s\ pier}$ ), pile-cap ( $y_{s\ pc}$ ) and pile group ( $y_{s\ pg}$ ). The calculation of the scour depth involves a step-by-step process, and the amount of scouring for the pile-cap ( $y_{s\ pc}$ ) and pile group ( $y_{s\ pg}$ ) need to take into account the impacts of the former (such as the pier part) to provide an updated soil covering height ( $h_1$ ,  $h_2$ , and  $h_3$ ), followed by re-calculating the equivalent flow velocity and equivalent pier width perpendicular to the flow for each part. After computing the three parts separately, the sum of the three parts would then yield the predicted scour depth, as shown in Eq. (3). Fig. 4 demonstrates that HEC-18 uses  $h_0$  to show the soil covering height;  $h_0$  is defined as the height of the pile cap above the bed at beginning of computation. This  $h_0$  is opposite from the Y value, and the relationship between the two can be expressed in Eq. (4).

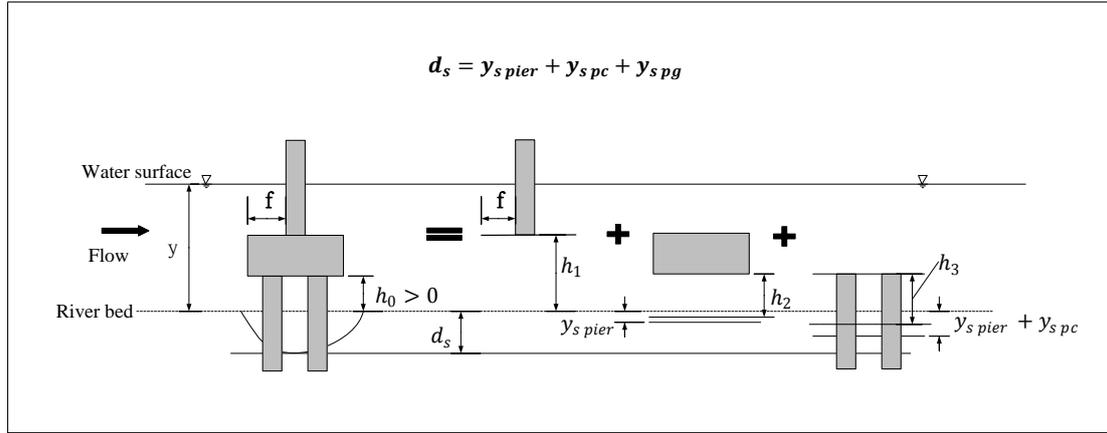


Fig. 4 Schematic view of HEC-18 non-uniform pier scour calculation

$$d_s = y_{s \text{ pier}} + y_{s \text{ pc}} + y_{s \text{ pg}} \quad (3)$$

$$h_0 = -(Y + T) \quad (4)$$

$$h_1 = h_0 + T$$

When HEC-18 is used, the  $Y$  values need to be utilized to determine the required formula, which can be divided into two scenarios,  $Y > 0$  and  $Y < 0$ . The case with  $Y > 0$  refers to (1), (2) and (3) in Fig. 5. The case with  $Y < 0$  refers to (4), (5), (6) and (7) in Fig. 5. When  $Y > 0$ , the bridge foundation is considered to be of a uniform pier, and the scour depth calculation is similar to Eq. (2), as shown in Eq. (5):

$$\frac{d_s}{y} = \left[ 2.0K_1K_2K_3K_w \left( \frac{b_c}{y} \right)^{0.65} \left( \frac{V_1}{\sqrt{gy}} \right)^{0.43} \right] \quad (5)$$

where  $V_1$  refers to the approach velocity used at the beginning of computations, and  $K_w$  refers to the correction factor for wide piers with shallow flow.

The case with  $Y < 0$ , it can be further classified into two scenarios,  $Y > -T$  or  $Y < -T$ . When  $Y > -T$ , referring to (4) in Fig. 5, the impact of the pile group does not need to be considered, but the scour depth needs to consider  $y_{s \text{ pier}}$  and  $y_{s \text{ pc}}$ . When computing the  $y_{s \text{ pier}}$ , because the pile-cap is already exposed in the water, it would have a protruding value ( $f$ , as shown in Fig. 4), and such an  $f$  value (distance between

front edge of pile-cap or footing and pie) would cause a shielding effect. HEC-18 uses  $K_{h\ pier}$  to consider such an effect, as expressed in Eq. (6). From Eq. (6), it can be learned that  $0 < K_{h\ pier} < 1$ , meaning that when  $Y > -T$ , the scour depth of the pier cannot be greater than the scour value of the uniform pier and it is not possible to be a negative value. At this time, the scour depth ( $y_{s\ pier}$ ) caused by the pier part can be calculated via Eq. (7). Whereas the scour depth ( $y_{s\ pc}$ ) caused by the pile-cap part can be calculated via Eq. (8).

$$K_{h\ pier} = \left( 0.04075 - 0.0669 \frac{f}{b_c} \right) - \left( 0.4271 - 0.0778 \frac{f}{b_c} \right) \left( \frac{h_1}{b_c} \right) + \left( 0.1615 - 0.0455 \frac{f}{b_c} \right) \left( \frac{h_1}{b_c} \right)^2 - \left( 0.0269 - 0.012 \frac{f}{b_c} \right) \left( \frac{h_1}{b_c} \right)^3 \quad (6)$$

$$\frac{y_{s\ pier}}{y} = K_{h\ pier} \left[ 2.0K_1K_2K_3K_w \left( \frac{b_c}{y} \right)^{0.65} \left( \frac{V_1}{\sqrt{gy}} \right)^{0.43} \right] \quad (7)$$

$$\frac{y_{s\ pc}}{y_f} = \left[ 2.0K_1K_2K_3K_w \left( \frac{b_{pc}}{y_f} \right)^{0.65} \left( \frac{V_f}{\sqrt{gy_f}} \right)^{0.43} \right] \quad (8)$$

where  $b_{pc}$  refers to the width of the original pile-cap,  $y_f = h_1 + y_{s\ pier} / 2$  is the distance from the bed to the top of the footing, and  $V_f$  is the average velocity in the flow zone below the top of the footing and can be calculated as follows.

$$\frac{V_f}{V_2} = \frac{\text{Ln} \left( 10.93 \frac{y_f}{K_s} + 1 \right)}{\text{Ln} \left( 10.93 \frac{y_2}{K_s} + 1 \right)} \quad (9)$$

where  $V_2 = V_1(y_1/y_2)$  is the average adjusted velocity in the vertical flow approaching the pier,  $V_1$  is the original approach velocity at the beginning of the computations,  $y_1$  is the original flow depth at the beginning of the computations before scouring,

$y_2 = y_1 + y_{s\ pier}/2 =$  the adjusted flow depth, and  $K_s$  is the grain roughness of the bed.

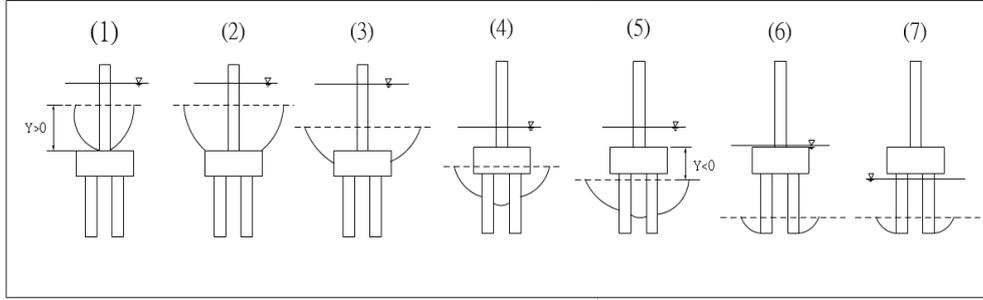


Fig. 5 Schematic view of various scour scenarios

For the case of  $Y < -T$ , referring to (5), (6) and (7) in Fig. 5, the scour depth calculation needs to consider  $y_{s\ pier}$ ,  $y_{s\ pc}$  and  $y_{s\ pg}$ . The calculation of  $y_{s\ pier}$  is similar to the above, and Eq. (7) is used for the calculation; as for the scour depth ( $y_{s\ pc}$ ) caused by the pile-cap part, the approach is similar to Eq. (8) but with modification because the pile-cap part is completely exposed in the water. Consequently, when  $Y < -T$ ,  $y_{s\ pc}$  is calculated via Eq. (10). As for the scour depth ( $y_{s\ pg}$ ) caused by the pile group part, it is calculated via Eq. (11).

$$\frac{y_{s\ pc}}{y_f} = \left[ 2.0K_1K_2K_3K_w \left( \frac{b_{pc}^*}{y_f} \right)^{0.65} \left( \frac{V_f}{\sqrt{gy_f}} \right)^{0.43} \right] \quad (10)$$

$$\frac{y_{s\ pg}}{y_3} = K_{h\ pg} \left[ 2.0K_1K_3 \left( \frac{b_{pg}^*}{y_3} \right)^{0.65} \left( \frac{V_3}{\sqrt{gy_3}} \right)^{0.43} \right] \quad (11)$$

where  $b_{pc}^*$  is the width of the equivalent pier and can be calculated using Eq. (12);

$y_3 = y_1 + y_{s\ pier}/2 + y_{s\ cap}/2 =$  the adjusted flow depth;  $K_{h\ pg}$  is the pile group height factor;  $b_{pg}^*$  is an equivalent pile group width that considers non-overlapping projected widths of piles, pile spacing, pile alignment and skewed or staggered pile groups; and  $V_3 = V_I(y_1/y_3) =$  the average adjusted velocity.

$$\frac{b_{pc}^*}{b_{pc}} = \text{Exp} \left[ -2.705 + 0.51 \text{Ln} \left( \frac{T}{y_2} \right) - 2.783 \left( \frac{h_2}{y_2} \right)^3 + 1.751 / \text{Exp} \left( \frac{h_2}{y_2} \right) \right] \quad (12)$$

where  $h_2 = h_0 + y_{s \text{ pier}}/2$ .

## **2.2 Existing method - the Method of Ataie-Ashtiani et al. (2010)**

HEC-18 considers the scour depth in the following three computation methods: (1) only the pier is considered, (2) the pier and pile-cap are considered, and (3) three parts of the pier, pile-cap and pile group are considered. Based on the experimental results, Ataie-Ashtiani B et al. (2010) claimed that the estimated scour depth of the pile-cap part is overly conservative and proposed a modification as expressed in Eq. (13).

$$d_s = K_A y_{s \text{ pier}} + K_B y_{s \text{ pc}} \quad (13)$$

where  $K_A$  refers to the correction coefficient of the pier part, as shown in Eq. (14).  $K_B$  refers to the correction coefficient of the pile-cap part, as shown in Eq. (15).

$$\begin{cases} K_A = 0.4 \left( 2.5 - \frac{f}{b_c} \right) & \text{if } \frac{f}{b_c} \leq 2.5 \\ K_A = 0 & \text{otherwise} \end{cases} \quad (14)$$

$$K_B = \left( \frac{b_{pc}}{y_{new}} \right)^{0.1} \quad (15)$$

where  $f$  refers to the distance between front edge of the pile-cap or footing and pie

and  $y_{new}$  refers to the corrected water depth  $= y + K_A y_{s \text{ pier}}$ .

## **2.3 Existing method – the Melville & Coleman’s approach**

Melville & Coleman (2000) proposed a prediction formula for the scour depth of complicated foundation, and the calculation method is expressed in Eq. (16).

$$d_s = K_{yb} K_s K_\theta K_I K_t K_d \quad (16)$$

where  $K_{yb}$  = water depth – bridge shape impact factor is as expressed in Eq. (17);  $K_s$  = pier shape correction factor is as expressed in Eq. (18);  $K_\theta$  = correction coefficient of the angle of attack of flow, as expressed in Eq. (19);  $K_I$  = flow intensity correction coefficient, as expressed in Eq. (20);  $K_t$  = time factor correction coefficient, as expressed in Eq. (21); and  $K_d$  = river bed material characteristic correction coefficient, as expressed in Eq. (22).

$$\begin{cases} K_{yb} = 2.4b_e & \frac{b_e}{y} < 0.7 \\ K_{yb} = 2\sqrt{yb_e} & 0.7 < \frac{b_e}{y} < 5 \\ K_{yb} = 4.5y & \frac{b_e}{y} > 5 \end{cases} \quad (17)$$

where  $b_e$  refers to the equivalent pier width perpendicular to the flow, and Eq. (17) suggests that when the flow depth  $y/b_e > 1.429$  (referring to  $b_e/y < 0.7$ ), the local scour depth of the pier is approximately 2.4 times the equivalent pier width ( $b_e$ ). When the flow depth  $y/b_e < 0.2$  (referring to  $b_e/y > 5$ ), the scour depth is only related to the water depth, which is approximately 4.5 times the water depth. When the water depth is between 0.2 and 1.429, the equivalent pier width ( $b_e$ ) and water depth both affect the local depth scour of the pier.

$$\begin{cases} K_s = 0.9 & \text{sharp nosed shape} \\ K_s = 1.0 & \text{circular, round nosed and group of cylinder} \\ K_s = 1.1 & \text{squared nosed shape} \end{cases} \quad (18)$$

Eq. (18) indicates that  $K_s$  is approximately between 0.9~1.1, and this range is close to the corrected range proposed by Raudkivi (1986). Raudkivi (1986) proposed that the impact of the pier shape on the pier scour depth was far less than the impact of the flow attack angle, and the range of the pier shape correction factor should be between 0.7~1.2.

$$K_{\theta} = \left( \frac{L}{b_e} \sin\theta + \cos\theta \right)^{0.65} \quad (19)$$

where  $L$  refers to the pier length and  $\theta$  refers to the angle of attack of the flow. In general, excluding circular column piers ( $K_{\theta} = 1$ ), the equivalent pier width ( $b_e$ ) would increase along with the increase in the angle of attack of the flow, as indicated in Eq. (19).

$$\begin{cases} K_t = \frac{V - (V_a - V_c)}{V_c} & \text{if } \frac{V - (V_a - V_c)}{V_c} < 1 \\ K_t = 1.0 & \text{otherwise} \end{cases} \quad (20)$$

where  $V_a$  refers to the non-uniform critical velocity of sediment movement.

$$K_t = \exp \left\{ -0.03 \left| \frac{V}{V_c} \ln \left( \frac{t}{t_e} \right) \right|^{1.6} \right\} \quad (21)$$

$$\begin{cases} K_d = 0.57 \log \left( 2.24 \frac{b_e}{d_{50}} \right) & \frac{b_e}{d_{50}} \leq 25 \\ K_d = 1 & \text{otherwise} \end{cases} \quad (22)$$

where  $d_{50}$  refers to the median grain size.

From the above computation process, it can be learned that in Melville & Coleman's approach, the equivalent pier width ( $b_e$ ) plays a key role. Additionally, when the water depth, river bed location and pier type are considered,  $b_e$  may be slightly different, which can mainly be classified into four cases (as shown in Fig. 6). The scenario of Case 1 ( $Y > b_{pc}$  where  $b_{pc}$  refers to the pile-cap width perpendicular to the flow) and the impacts of the pile-cap and pile group can be ignored. In this case,  $b_e = b_c$ . The values of  $b_e$  for Case 2 ( $Y \leq b_{pc}$  and  $Y > 0$ ) and Case 3 ( $Y \leq 0$  and  $-Y < y$ ) need to consider the width perpendicular to the flow for both the pier and pile-cap. During the calculation of the scour depth in Case 4,  $b_e$  is assumed to be  $b_{pc}$ . The four types of cases, after organization, can be expressed in Eq. (23).

$$\left\{ \begin{array}{l} b_e = b_c \\ b_e = b \left( \frac{y+Y}{y+b_{pc}} \right) + b_{pc} \left( \frac{b_{pc}-Y}{y+b_{pc}} \right) \\ b_e = b \left( \frac{y+Y}{y+b_{pc}} \right) + b_{pc} \left( \frac{b_{pc}-Y}{y+b_{pc}} \right) \\ b_e = b_{pc} \end{array} \right. \begin{array}{l} Y > b_{pc} \\ b_{pc} \geq Y \geq 0 \\ 0 \geq Y \geq -y \\ -Y > y \end{array} \quad (23)$$

From Eq. (23), it can be learned that for Cases 2 and 3 in Melville & Coleman's approach, the calculations of  $b_e$  are identical.

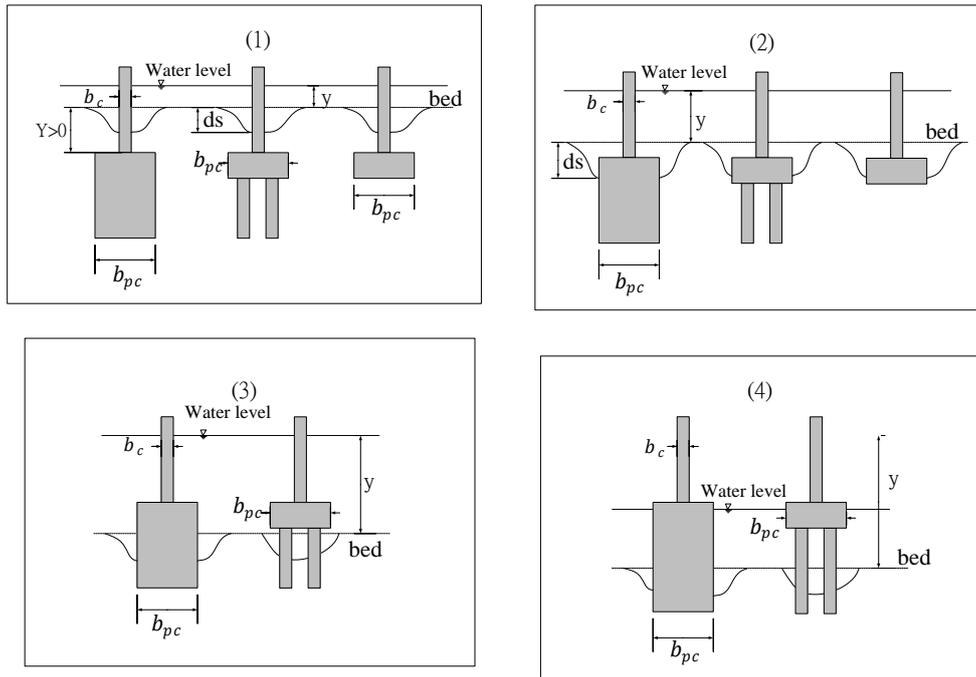


Fig. 6 Calculation of equivalent width perpendicular to the flow under four different types of scour cases.

## **2.4 The proposed calculations for the scour depth of a bridge with a complex pier**

Two alternative approaches for calculating the scour depth are proposed. The first method uses the machine learning theory, and the second method uses the sequential quadratic programming to find the optimal coefficients in the proposed formula, as described below:

### **2.4.1 Using machine learning technique**

In addition to the existing formulae introduced earlier, Artificial Neural Networks (ANNs) or Support Vector Machines (SVMs) are potential tools that can be used to build a prediction model of scour depth. ANNs have been successfully applied to many civil engineering problems, such as predicting the bearing capacity of strip footing (Kuo et al., 2009), analyzing the slope stability analysis (Cho, 2009), predicting the rock fragmentation due to blasting (Bahrami et al., 2011), predicting the groutability of microfine cements in permeation grouting (Liao et al. 2011) and predicting the scour depth (Hosseini et al. 2016 and Lashkar-Ara et al. 2016). The SVM is another useful technique for data classification and regression. Constructing a SVM is often considered to be easier than building an ANN model. SVMs also have been applied to many engineering problems, such as predicting the blast-induced ground vibration (Khandelwal, 2011), identifying the lateral flow occurrence (Lee and Kim, 2010), predicting the side weir discharge coefficient (Azamathulla et al. 2016), forecasting head loss on cascade weir (Haghiabi et al. 2016) and detecting the rusted area in a steel bridge (Liao and Lee, 2016). Although the SVM has been recognized a powerful tool, only few of researches has investigated its suitability in scour depth prediction (). Thus, SVM is chosen as one of the proposed approaches to calculate the scour depth for a bridge with a complex pier.

A standard SVM, as described in Eq. (24), solves a nonlinear classification problem by means of convex quadratic programs (QP).

$$\begin{aligned} & \underset{w,b,\xi}{\text{minimize}} \quad \frac{1}{2} w^T w + c \sum_{k=1}^N \xi_k \\ & \text{Subject to} \quad \begin{cases} y_k (w^T K(x_i) + b) \geq 1 - \xi_k \\ \xi_k \geq 0, \quad i = 1, 2, \dots, N \end{cases} \end{aligned} \quad (24)$$

where  $w$  is a normal vector to the hyper-plane;  $c$  is a real positive constant; and  $\xi_k$  is the slack variable. If  $\xi_k > 1$ , the  $k$ -th inequality becomes violated compared to the

inequality from the linearly separable case.  $y_k$  is the class;  $[w^T K(x_i) + b]$  is the classifier;  $N$  is the number of data; and  $K$  is the kernel function. In the current study, the Gaussian radial basis function (RBF) kernel is used, as shown in Eq. (25).

$$K(X, X_i) = e^{-\sigma(\|X - X_i\|)^2} \quad (25)$$

where  $X$  is the input vector,  $\sigma$  is the kernel function parameter; and  $X_i$  are the support vectors. LS-SVM (Suykens et al. 2002), instead of solving the QP problem, solves a set of linear equations by modifying the standard SVM, as described in Eq. (26).

$$\begin{aligned} \min \quad & \frac{1}{2} w^T w + \frac{\gamma}{2} \sum_{k=1}^N e_k^2 \\ \text{s.t.} \quad & y_k (w \cdot K(x_k) + b) = 1 - e_k, \quad k=1, \dots, n \end{aligned} \quad (26)$$

where  $\gamma$  is a constant number and  $\varepsilon$  is the error variable. Compared to the standard SVM, there are two modifications leading to solving a set of linear equations. First, instead of inequality constraints, the LS-SVM uses equality constraints. Second, the error variable is a squared loss function.

Because the input data play an important role in the LS-SVM, the selections of input parameters are described below. Two LS-SVM models are developed. The inputs of the first and second models are approximately considered as HEC-RAS-based and Melville & Coleman-based LS-SVMs. Based on the aforementioned introduction and Buckingham  $\pi$  theorem, the results of the dimensional analysis for the HEC-RAS and Melville & Coleman approaches can be described as shown in Eqs. (27) and (28).

$$\frac{d_s}{b_c} = f \left( \frac{V}{\sqrt{gy}}, \frac{y}{b_c}, \frac{D_{50}}{b_c}, \sigma_s, \frac{V}{V_c}, \frac{b_{pc}}{b_c}, \frac{b_{pg}}{b_c}, \frac{Y}{b_c}, \frac{f}{b_c}, \frac{L}{b_c} \right) \quad (27)$$

$$\frac{d_s}{b_e} = f \left( \frac{V}{V_c}, \frac{y}{b_c}, \frac{D_{50}}{b_e}, \sigma_s, \frac{Y}{b_e}, \frac{L}{b_e} \right) \quad (28)$$

Although the two sets of input factors are similar, differences between them are

noticeable. For example, the basic dimensions of HEC-RAS and Melville & Coleman are  $b_c$  and  $b_e$ , respectively. Furthermore, the widths of the pile-cap and pile groups are explicitly considered in the HEC-RAS-based LS-SVM model. Please note that factors displayed in Eqs.(27) and (28) only include the available factors in the collected experiment data. Table 1 displays the data range used for developing LS-SVM in the current study.

Table 1 Data range used for developing LS-SVM

	$y$	$b_c$	$b_{pc}$	$L_u$	$Y$	$V/V_c$	$d_{50}$	$\sigma_g$	$d_s$
Max.	0.6	0.1524	0.3694	2.78	0.205	1.183206	0.001	1.3	0.338328
Min.	0	0	0	0	-0.67	0	0	0	0

#### **2.4.2 Using formula-based approach**

It is recognized that neither ANNs nor SVMs provide a specific formula for engineers, although they have the potential to deliver a promising prediction (Huang et al., 2013). The concept and operation of an ANN or an SVM model is a black box for some practical engineers. On the other hand, an explicit formula is often used in civil/hydraulic engineering. Thus, one of the goals in this study is aimed to develop a formula with a similar format to that of existing formulae that provides an analogous evaluation procedure but with higher accuracy for predicting scour depth in a bridge with a complex pier. Details are provided below.

Although the concept of superposition used in HEC-18 is straightforward, updating several parameters in different cases complicates the prediction process (Section 2.1). Melville & Coleman (2000) treats the pier as a single element resulting in a single formula, which is more suitable to the engineering application in practice.

Therefore, the formula in Melville & Coleman (2000) is used as the basis to develop a new prediction formula. Eq. (23) describes the calculation of  $b_e$ ; it is seen that  $b_e$  is basically interpolated using the two values of  $b_c$  and  $b_{pc}$ , as shown in Eq. (29).

$$b_e = A \times b_c + B \times b_{pc}, \text{ where } A + B = 1 \quad (29)$$

where  $A$  and  $B$  are the weights for  $b_c$  and  $b_{pc}$ , respectively, and the sum of the two weights is 1. According to the suggestion of Melville & Coleman (2000),  $A$  and  $B$  are functions of the flow depth ( $y$ ), the level of the top surface of pile cap below surrounding bed level ( $Y$ ) and the pile-cap width perpendicular to the flow ( $b_{pc}$ ). Similarly, the optimization technique adopted here is used to obtain the function content of  $A$  and  $B$ , as described in Eq. (30).

$$b_e = \left( \frac{x_1 * y + x_2 * Y}{x_3 * y + x_4 * b_{pc}} \right) b_c + \left( \frac{x_5 * b_{pc} - x_6 * Y}{x_7 * y + x_8 * b_{pc}} \right) b_{pc} \quad (30)$$

where  $x_i$  refers to the coefficient to be determined. In addition to modifying the formula of  $b_e$ , from Eq. (23), it can be learned that the accuracy of the prediction of the scour depth also depends on the correct classification. For example, Eq. (23) classifies the formula into three types based on the value of  $Y$  (formulae for cases 2 and 3 are identical). However, according to the experimental results (e.g., Melville and Raudkivi, 1996), as shown in Fig. 7, it is obvious that the change in the scour depth near  $Y=0$  is extremely high such that it is not appropriate to use one identical formula (referring to Eq. (23)) for the calculations of the  $b_e$  values at two areas of  $Y$  that are greater than or smaller than 0. Therefore, calculation of  $b_e$  is classified into four different cases, as shown in Eq. (31), and the corresponding functions, such as  $f$ ,  $g$ ,  $h$  and  $k$ , are determined through the optimization method.

$$\begin{cases} b_e = f(y, Y, b_c, b_{pc}) & Y > 2.4b_c \\ b_e = g(y, Y, b_c, b_{pc}) & 2.4b_c \geq Y \geq 0 \\ b_e = h(y, Y, b_c, b_{pc}) & 0 \geq Y \geq -y \\ b_e = k(y, Y, b_c, b_{pc}) & -Y > y \end{cases} \quad (31)$$

where the mathematical formulation of the optimization problem is described as follows.

$$\begin{aligned} \text{Min.} \quad & D_s - K_{yb} K_s K_\theta K_I K_t K_d \\ \text{s.t.} \quad & x_i, \quad i = 1, 2, \dots, 8 \end{aligned} \quad (32)$$

where  $D_s$  refers to the scour depth obtained from the experiment,  $K_{yb} K_s K_\theta K_I K_t K_d$  is the function of  $b_e$ , and the calculation of  $b_e$  is described in Eq. (30).

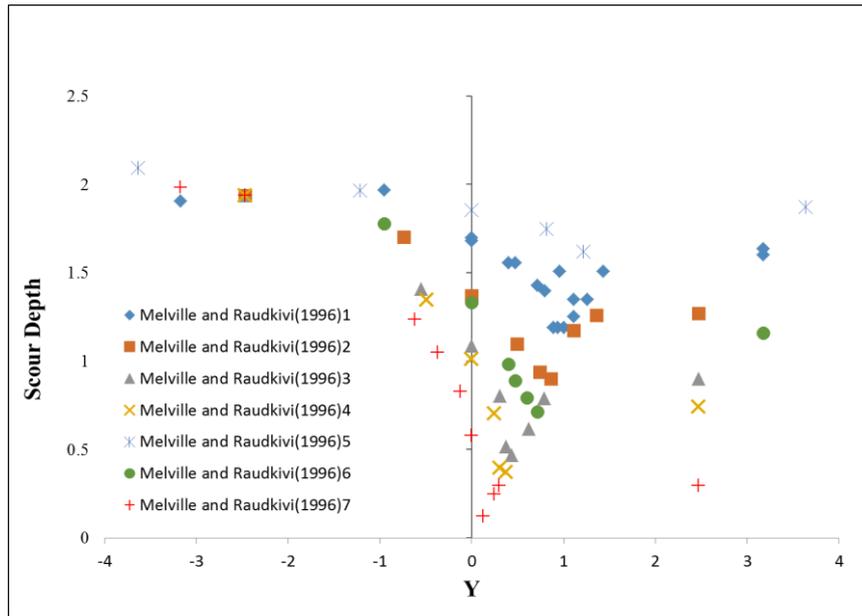


Fig. 7 Relationship between the scour depth ( $d_s$ ) and soil covering depth ( $Y$ )

### **3.0 Results and discussions**

#### **3.1 Data and prediction results using existing formulae**

The scour depth experimental data collected in this study are shown in Table 2, including a total of 175 experimental data entries (comprising the four data entries of this research). The methods of Melville & Coleman (2000), HEC-18 (2012),

Ataie-Ashtiani et al. (2010) and the proposed approaches including LS-SVM and formula-based method are used to perform the scour depth calculation. The calculated results are then compared with the experimental data to evaluate the accuracy of each method.

Table 2 Data source and information for the collected experiments

Sources	$y/b_c$	$b_c/D_{50}$	$V/V_c$	$Y/b_c$	$f/b_c$	$T/b_c$	$b_{pg}/b_c$
Sheppard & Renna (2005)	2	152	1.18	-0.2-1	0.1-0.5	0.2	-
Ataie-Ashtiani et al. (2010)	3.3-7.1	37-70	0.72-0.85	-8.2-3.2	0.42-0.68	1-1.45	0.38-0.73
Melville & Raudkivi (1996)	4.4-20	12-188	$\approx 1$	-20-2.5	0.11-3.55	-	-
Coleman (2005)	3.3-7.1	37-70	0.72-0.85	-8.2-3.2	0.42-0.68	1-1.45	0.38-0.73
Present study	3.57-4.18	76.5	0.53-0.91	0	0.454	0.954	0.68

Figs. 8-10 are the comparison charts of the analysis and experiment results conducted on the 175 data entries using the Melville & Coleman (2000), HEC-18 (2012) and Ataie-Ashtiani et al. (2010) methods; wherein, the X axis refers to the actual scour and Y axis refers to the predicted scour depth. The prediction is considered accurate when the data points in Figs. 8-10 are on the reference line. Table 3 shows the accuracy evaluations on the aforementioned three methods. In Figs. 8-10 and Table 3, it can be seen that the prediction accuracy of the method of Melville & Coleman (2000) is relatively inaccurate. If the evaluation standard specified by Lewis (1982) is used (as shown in Table 4), the three calculation methods are all undesirable and are rated as inadequate.

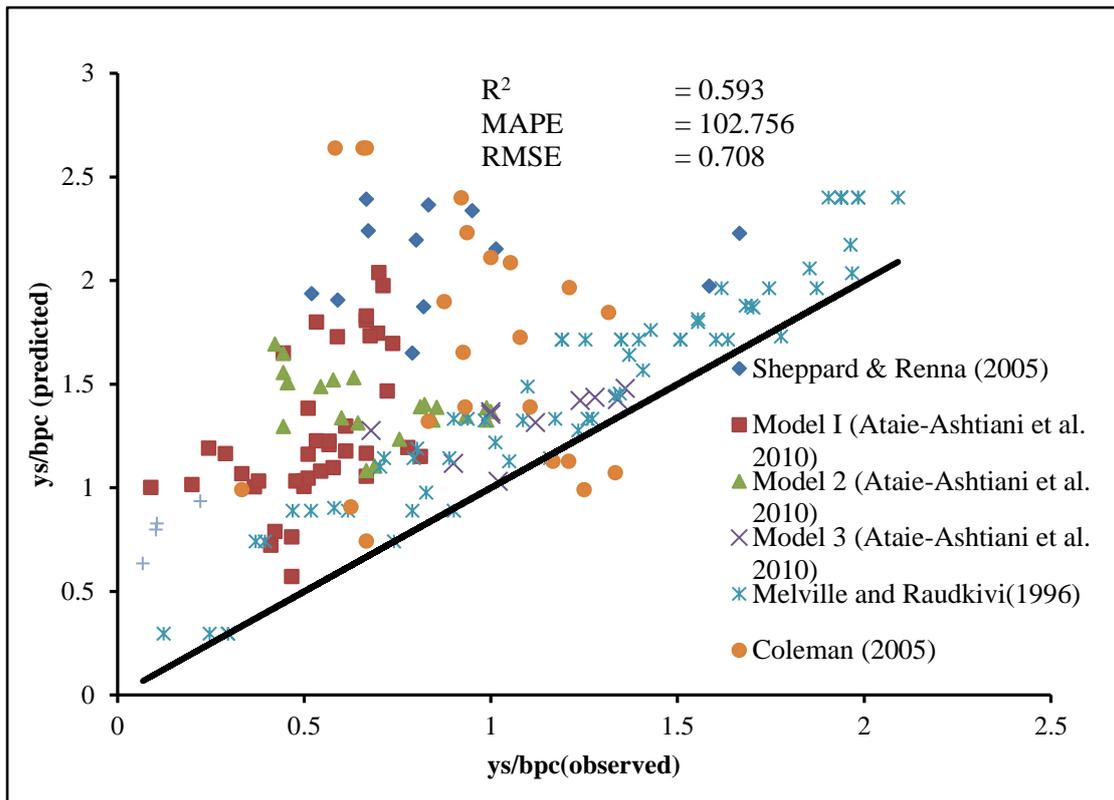


Fig. 8 Comparison between the prediction scour values (Melville & Coleman, 2000) and actual scour values

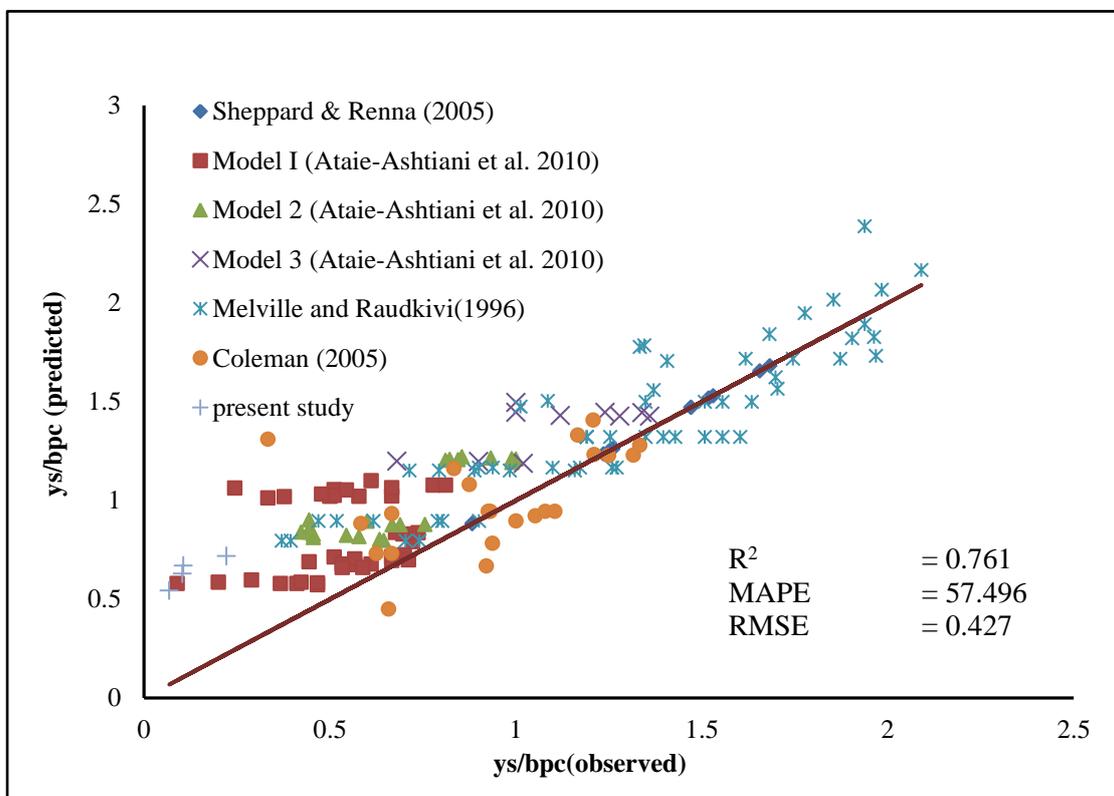


Fig. 9 Comparison between the prediction scour values (HEC-18) and actual scour values

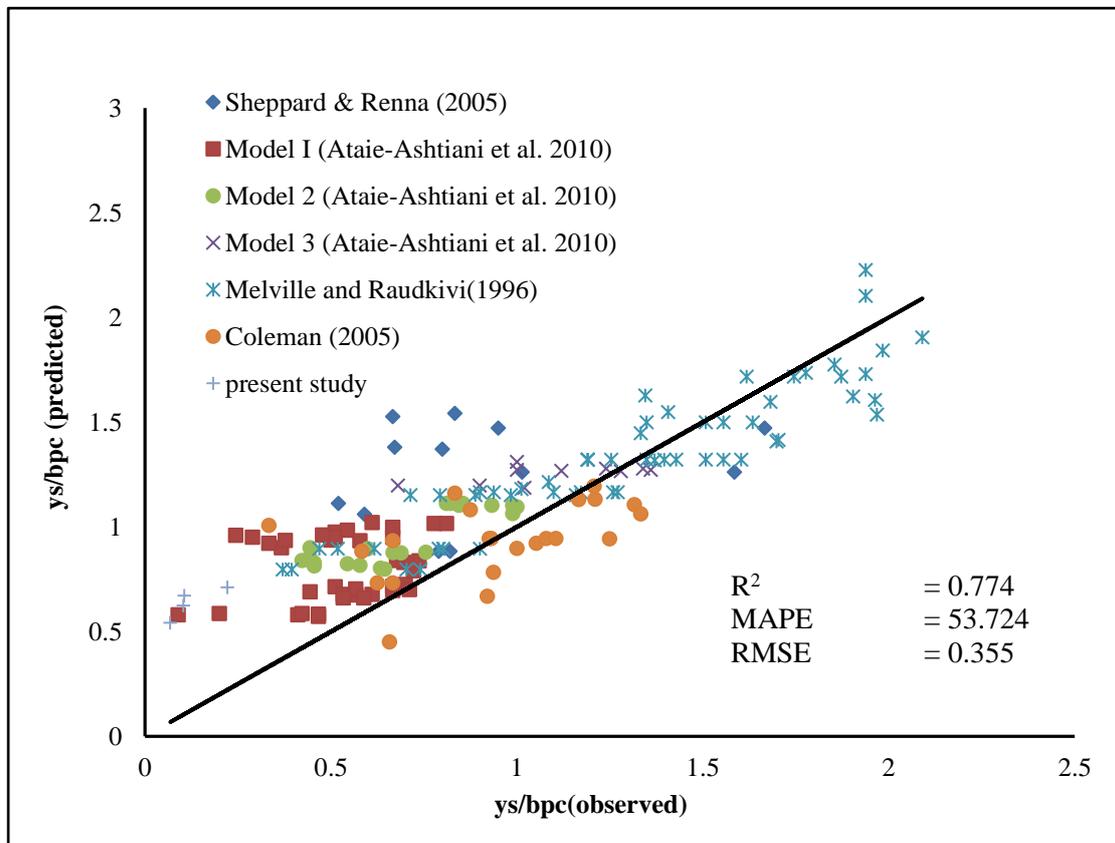


Fig. 10 Comparison between the prediction scour values (Ataie-Ashtiani et al. 2010) and actual scour values

Table 3 Error evaluations of scour formulae

Calculation method	MAPE	RMSE
Melville& Coleman (2000)	102.7564	0.707848
HEC-18	57.4965	0.427707
Ataie-Ashtiani et al. (2010)	53.7245	0.354607

Table 4 MAPE evaluation ranking table

MAPE (%)	Evaluation ranking
< 10	Most optimal
10 ~ 20	Excellent

20 ~ 50	Fair
> 50	Inadequate

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### **3.2 Discussions for the existing formulae**

From Fig. 8 and Table 3, it can be learned that for the method proposed by Melville et al. (2000), among the 175 data entries, only 8 data entries appear at the right side of the reference line, indicating that the overall prediction is too conservative. From Eq. (16), it is learned that the formula proposed by Melville et al. (2000) is mainly affected by six factors. Table 5 lists the numerical ranges of these factors that should be used as a basis for studying the source of the error. As shown in Table 5, the values of  $K_s$ ,  $K_\theta$  and  $K_d$  nearly have no impact on the predicted scour depth; for these, because no experiments consider the angle of attack of flow, the values of  $K_\theta$  are all equal to 1. Consequently, the main source of error could be  $K_{yb}$ ,  $K_l$ , and  $K_t$ . Wherein  $K_t$  is especially problematic because most of the experiments have extended the testing time to reduce the impact of time on the scour depth, and only an experimental value in all of the data of  $K_l$  is smaller than 0.7, while more than half of the values of the experimental result of  $K_l$  are equivalent to 1. Because of this,  $K_l$  and  $K_t$  should not be factors that cause  $\text{MAPE} > 100$ . In view of the above discussion, the key factor for error is  $K_{yb}$ . From Eq. (16), it is known that  $K_{yb}$  is greatly affected by the equivalent width perpendicular to the flow ( $b_e$ ). Therefore, the formulation of  $b_e$ , as described in Section 3.2., is revised to enhance the prediction accuracy, as shown in Section 4.3.

Table 5 Numerical ranges of the six impact factors of  
(Melville& Coleman, 2000)

Impact factor	Numerical range
$K_{yb}$	0.024 ~ 0.54
$K_s$	1.0 ~ 1.1

$K_{\theta}$	1
$K_I$	0.53 ~ 1
$K_t$	0.76 ~ 1
$K_d$	1

Fig. 9 and Table 3 show that the scour depth obtained from HEC-18 is of a smaller error compared to that of Melville et al. (2000). However, according to Table 4, the prediction result is inadequate.  $K_{h\ pier}$  is a possible factor that is described as follows. Parola et al. (1996) suggested that a rectangular pile-cap length would reduce the local scour depth. When the extension length of the pile-cap toward the upstream direction ( $f$ , as shown in Fig. 4) is approximately 2.3-2.5 of  $b_c$  (pier width perpendicular to the flow), it is able to effectively reduce the scour depth. In the collected data, the ratio of  $f/b_c$  is between 0.2~7.1. When the ratio  $f/b_c$  exceeds 2.5, the trend of  $K_{h\ pier}$  (as shown in Fig. 11) does not follow the direction suggested by HEC-18 (Section 2.1,  $0 < K_{h\ pier} < 1$ , and it cannot be a negative value). When  $f/b_c$  is greater than 2.5, the applicability of the  $K_{h\ pier}$ , which was suggested by HEC-18 to take the effect of  $f/b_c$ , is questionable.

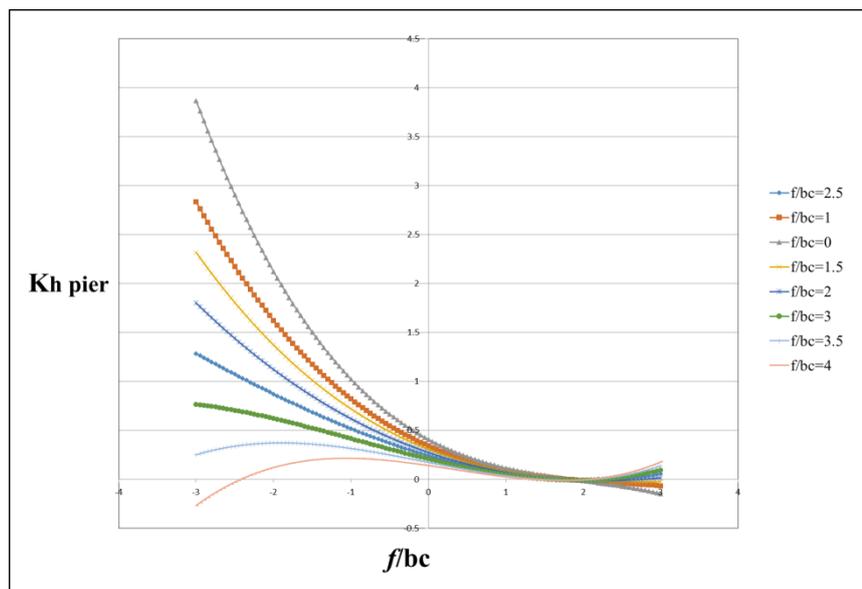


Fig. 11 Relationship between  $f/b_c$  and  $K_{h \text{ pier}}$

### **3.3 Prediction results using the proposed methods**

Table 6 shows results of using LS-SVM to predict the scour depth. The common 5-fold method is used to prevent overfitting of the data. In addition, to decrease overfitting, a total of 9 analyses of 5-fold are conducted, meaning that for each LS-SVM model, there are a total 45 prediction results. Table 6 shows the average value and coefficient of variation (COV) of the prediction result. According to Table 6, the results of MAPE are generally 25~27, which are within the reasonable range (Table 4). Compared to the existing formulae used, LS-SVM increases the accuracy by approximately 2~4 times. Regardless of whether it is MAPE or  $R^2$ , the two LS-SVM prediction results do not significantly differ from each other. By using MAPE as an example, there is only a difference of approximately 2%, which means that the two sets of different input parameters have relatively small impact on the LS-SVM results as well as that the input parameters considered by Melville & Coleman (2000) should yield prediction results that are similar to the ones of HEC-18. Because engineers often prefer the formula-based approach over a machine learning method, this study proposed an alternative approach and the results are described as follows. Table 7 provided an error comparison using different kernel functions, indicating that RBF is a better selection than the linear kernel. Please note that other kernel functions are available for the use in SVM, their performance investigation is important but is beyond the scope of current study.

Table 6 Accuracies of using LS-SVM to predict the scour depth

Analysis method	MAPE (%)	$R^2$
Melville& Coleman-based LS-SVM	27.2 (0.31)*	0.79 (0.19)*
HEC-18-based LS-SVM	25.1 (0.28)*	0.84 (0.06)*

\*the number inside the brackets refers to COV

Table 7 Accuracies of Melville& Coleman-based LS-SVM for different kernel functions

Kernel function	MAPE (%)	R <sup>2</sup>
RBF	27.2 (0.31)*	0.79 (0.19)*
Linear	34.6 (0.02)*	0.62 (0.02)*

\*the number inside the brackets refers to COV

The sequential quadratic programming from the MATLAB toolbox is used to solve the optimization problem described in Eq. (32). The objective of the optimization is to find 8 coefficients of the functions of  $f$ ,  $g$ ,  $h$  and  $k$  in Eq. (31); wherein when  $Y > 2.4b_c$  it is typically recognized so that it is not scoured to the location of the pile-cap and so that the influence of pile-cap and pile groups can be ignored. Therefore, under such conditions, optimization is not performed, indicating that  $f = b_c$  and  $b_e = b_c$ . Optimization results are described in Eqs. (33), (34) and (35).

$$\begin{cases} b_e = f(y, Y, b_c, b_{pc}) & Y > 2.4b_c \\ b_e = g(y, Y, b_c, b_{pc}) & 2.4b_c \geq Y \geq 0 \\ b_e = h(y, Y, b_c, b_{pc}) & 0 \geq Y \geq -y \\ b_e = k(y, Y, b_c, b_{pc}) & -Y > y \end{cases} \quad (31)$$

$$b_e = g(y, Y, b_c, b_{pc}) = \left( \frac{0.80y + 0.31Y}{0.83y + 1.00b_{pc}} \right) b_c + \left( \frac{0.02b_{pc} - 0.07Y}{0.75y + 0.56b_{pc}} \right) b_{pc} \quad (33)$$

$$b_e = h(y, Y, b_c, b_{pc}) = \left( \frac{0.22y + 0.38Y}{0.16y + 0.31b_{pc}} \right) b_c + \left( \frac{0.05b_{pc} - 0.16Y}{-0.03y + 0.57b_{pc}} \right) b_{pc} \quad (34)$$

$$b_e = k(y, Y, b_c, b_{pc}) = \left( \frac{0.42y + 0.11Y}{0.24y + 0.90b_{pc}} \right) b_c + \left( \frac{0.11b_{pc} - 0.08Y}{0.20y + 1.00b_{pc}} \right) b_{pc} \quad (35)$$

Table 8 shows the prediction result of the proposed formula-based approach. In general, the result greatly improves the accuracy of the Melville & Coleman (2000)

formula. The proposed method is able to simultaneously satisfy the requirements of accuracy and simplicity. The proposed formula has the advantages of being conceptually consistent with the observed scour behaviors and provides a solid scour depth prediction, which is an important and critical step in the bridge safety evaluation if floods are considered.

Table 8 Accuracies of the proposed formula-based approach

Soil covering depth	MAPE
(1) $Y > 2.4b_c$	5.1
(2) $2.4b_c > Y \geq 0$	30.4
(3) $0 > Y > -y$	34.2
(4) $Y \leq -y$	24.8
Average	28.9

#### **4.0 Conclusions**

Taiwan is an elongated island with many rivers, and river bridges have become important traffic links. The pile group foundation is one of the main structures for bridges in Taiwan. However, establishing a practical approach to estimate the local scour depth for a pile group foundation has not drawn many attentions. Thus, a prediction formula of a relatively simple form with sufficient accuracy is preferred. To fulfill such target, experimental data of a total of 175 entries are collected to investigate accuracies of three available scour formulae, machine learning method and the proposed formula. Based on the analyses results, several important conclusions can be drawn as below.

1. The predictions of all three existing formulae do not provide satisfactory outcomes.

2. For the Melville & Coleman (2000) method, the equivalent width

perpendicular to the flow ( $b_e$ ) is a major source of error. On the other hand,  $K_{h\ pier}$ , suggested by HEC-18, fails to act like  $f/b_c$  when it is greater than 2.5.

3. The two LS-SVM models significantly improve the prediction performance and do not greatly vary from each other indicating that all scour contributing factors have been included in the two existing formulae of Melville & Coleman (2000) and HEC-18.

4. The results of the proposed formula, adopting the concept of the equivalent width, is able to significantly increase the prediction accuracy for most of conditions compared to the existing formulae.

5. The proposed formula-based approach has a similar format to that of the existing formulae, providing an analogous evaluation procedure without increasing the application difficulties.

Although there is still room for further refinement when the flow depth is lower than the pile-cap top for the proposed formula, the proposed formula has the advantages of being conceptually consistent with the observed scour behaviors and provides a satisfied scour depth prediction. Please note that the formula proposed was built using the collected experimental data, further validation is needed to generalize its use in future applications.

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## Appendix A: the dataset used in this study

y	$b_c$	$b_{pc}$	Lu	Y	V/V <sub>c</sub>	d50 (10 <sup>-3</sup> )	$\sigma_g$	d <sub>s</sub>
0.20	0.05	0.08	0.04	0.11	1.00	0.80	1.00	0.10
0.20	0.03	0.08	0.05	0.04	1.00	0.80	1.00	0.04
0.60	0.03	0.12	0.05	0.03	0.75	0.47	1.30	0.08
0.30	0.15	0.21	0.06	0.15	1.18	1.00	1.00	0.34
0.20	0.01	0.08	0.07	0.02	1.00	0.80	1.00	0.02
0.20	0.03	0.08	0.06	0.20	1.00	0.80	1.00	0.06
0.20	0.05	0.06	0.01	0.05	1.00	0.80	1.00	0.10
0.60	0.03	0.12	0.05	-0.12	0.75	0.47	1.30	0.10
0.20	0.05	0.08	0.04	-0.20	1.00	0.80	1.00	0.16
0.15	0.02	0.09	0.15	-0.03	0.76	0.60	1.20	0.07
0.30	0.15	0.21	0.06	-0.03	1.18	1.00	1.00	0.14
0.20	0.05	0.06	0.01	-0.20	1.00	0.80	1.00	0.12
0.20	0.03	0.08	0.06	-0.04	1.00	0.80	1.00	0.11
0.22	0.06	0.37	2.78	0.00	0.82	0.80	1.00	0.04
0.33	0.10	0.19	0.05	-0.41	0.83	0.47	1.30	0.13
0.60	0.03	0.12	0.05	-0.67	0.75	0.47	1.30	0.07
0.14	0.02	0.09	0.15	-0.15	0.80	0.60	1.20	0.06
0.14	0.03	0.05	0.01	0.05	0.79	0.06	1.20	0.05
0.15	0.04	0.09	0.03	0.01	0.75	0.60	1.20	0.07
0.20	0.05	0.06	0.02	0.20	1.00	0.24	1.00	0.10
0.60	0.03	0.12	0.05	0.09	0.75	0.47	1.30	0.08
0.20	0.03	0.06	0.03	0.04	1.00	0.80	1.00	0.05
0.20	0.03	0.06	0.03	0.03	1.00	0.80	1.00	0.06
0.15	0.04	0.09	0.03	-0.02	0.78	0.60	1.20	0.08
0.15	0.02	0.09	0.15	0.00	0.84	0.60	1.20	0.02
0.30	0.15	0.30	0.15	0.00	1.18	1.00	1.00	0.18
0.15	0.03	0.05	0.01	0.00	0.79	0.06	1.20	0.05
0.30	0.15	0.21	0.06	0.00	1.18	1.00	1.00	0.17
0.20	0.05	0.06	0.02	-0.20	1.00	0.24	1.00	0.12
0.15	0.04	0.09	0.03	-0.02	0.78	0.60	1.20	0.09
0.15	0.04	0.09	0.03	-0.10	0.77	0.60	1.20	0.04
0.14	0.02	0.09	0.15	-0.18	0.76	0.60	1.20	0.05
0.15	0.04	0.09	0.03	-0.09	0.76	0.60	1.20	0.04
0.15	0.02	0.09	0.15	-0.11	0.79	0.60	1.20	0.06
0.30	0.15	0.30	0.15	0.03	1.18	1.00	1.00	0.25

0.20	0.03	0.08	0.06	0.02	1.00	0.80	1.00	0.06
0.20	0.05	0.06	0.02	0.09	1.00	0.24	1.00	0.10
0.33	0.10	0.19	0.05	0.21	0.83	0.47	1.30	0.21
0.20	0.05	0.06	0.02	0.06	1.00	0.80	1.00	0.10
0.15	0.04	0.09	0.03	0.04	0.75	0.60	1.20	0.06
0.20	0.05	0.06	0.02	0.08	1.00	0.24	1.00	0.09
0.15	0.04	0.09	0.03	-0.03	0.75	0.60	1.20	0.07
0.15	0.02	0.09	0.15	-0.01	0.76	0.60	1.20	0.04
0.26	0.06	0.37	2.78	0.00	0.70	0.80	1.00	0.04
0.30	0.15	0.30	0.15	-0.03	1.18	1.00	1.00	0.16
0.20	0.03	0.08	0.05	-0.05	1.00	0.80	1.00	0.11
0.15	0.02	0.09	0.15	-0.01	0.74	0.60	1.20	0.04
0.20	0.03	0.06	0.03	-0.20	1.00	0.80	1.00	0.13
0.20	0.01	0.08	0.07	-0.01	1.00	0.80	1.00	0.07
0.15	0.02	0.09	0.15	-0.16	0.72	0.60	1.20	0.04
0.15	0.02	0.09	0.15	-0.11	0.77	0.60	1.20	0.07
0.14	0.02	0.09	0.15	-0.16	0.79	0.60	1.20	0.06
0.30	0.15	0.30	0.15	0.15	1.18	1.00	1.00	0.24
0.20	0.03	0.08	0.05	0.03	1.00	0.80	1.00	0.04
0.15	0.02	0.09	0.15	0.01	0.79	0.60	1.20	0.02
0.20	0.03	0.06	0.03	0.20	1.00	0.80	1.00	0.07
0.33	0.10	0.19	0.05	0.16	0.83	0.47	1.30	0.18
0.15	0.02	0.09	0.15	0.04	0.79	0.60	1.20	0.04
0.14	0.04	0.09	0.03	0.01	0.80	0.60	1.20	0.05
0.15	0.04	0.09	0.03	-0.01	0.79	0.60	1.20	0.08
0.20	0.05	0.06	0.02	-0.20	1.00	0.80	1.00	0.13
0.15	0.03	0.05	0.01	-0.01	0.77	0.06	1.20	0.05
0.20	0.05	0.08	0.04	-0.06	1.00	0.80	1.00	0.14
0.60	0.03	0.12	0.05	-0.05	0.84	0.47	1.30	0.15
0.15	0.03	0.05	0.01	-0.03	0.78	0.06	1.20	0.07
0.15	0.02	0.09	0.15	-0.01	0.74	0.60	1.20	0.05
0.20	0.03	0.06	0.03	-0.06	1.00	0.80	1.00	0.11
0.14	0.04	0.09	0.03	-0.10	0.78	0.60	1.20	0.04
0.15	0.04	0.09	0.03	-0.07	0.77	0.60	1.20	0.04
0.15	0.04	0.09	0.03	-0.06	0.74	0.60	1.20	0.06
0.20	0.05	0.06	0.02	0.07	1.00	0.24	1.00	0.08
0.20	0.05	0.06	0.01	0.07	1.00	0.80	1.00	0.09
0.13	0.02	0.09	0.15	0.06	0.80	0.60	1.20	0.04

0.20	0.05	0.06	0.02	0.20	1.00	0.80	1.00	0.10
0.20	0.03	0.08	0.05	0.05	1.00	0.80	1.00	0.05
0.20	0.05	0.08	0.04	0.07	1.00	0.80	1.00	0.07
0.15	0.02	0.09	0.15	-0.01	0.79	0.60	1.20	0.05
0.20	0.03	0.08	0.05	-0.20	1.00	0.80	1.00	0.16
0.20	0.05	0.06	0.02	0.00	1.00	0.80	1.00	0.11
0.15	0.02	0.09	0.15	-0.02	0.85	0.60	1.20	0.06
0.20	0.03	0.08	0.06	-0.20	1.00	0.80	1.00	0.16
0.15	0.04	0.09	0.03	-0.02	0.77	0.60	1.20	0.08
0.20	0.03	0.08	0.06	0.00	1.00	0.80	1.00	0.08
0.26	0.06	0.37	2.78	0.00	0.91	0.80	1.00	0.08
0.15	0.02	0.09	0.15	-0.04	0.75	0.60	1.20	0.05
0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
0.33	0.10	0.19	0.05	-0.11	0.83	0.47	1.30	0.19
0.20	0.05	0.06	0.02	0.05	1.00	0.24	1.00	0.09
0.33	0.10	0.19	0.05	0.05	0.83	0.47	1.30	0.21
0.15	0.03	0.05	0.01	0.01	0.77	0.06	1.20	0.03
0.20	0.03	0.08	0.05	0.03	1.00	0.80	1.00	0.07
0.20	0.05	0.08	0.04	0.20	1.00	0.80	1.00	0.10
0.15	0.03	0.05	0.01	0.06	0.76	0.06	1.20	0.05
0.15	0.04	0.09	0.03	0.02	0.80	0.60	1.20	0.04
0.15	0.02	0.09	0.15	-0.02	0.72	0.60	1.20	0.06
0.30	0.15	0.18	0.03	-0.03	1.18	1.00	1.00	0.12
0.20	0.05	0.06	0.02	0.00	1.00	0.24	1.00	0.11
0.20	0.01	0.08	0.07	-0.20	1.00	0.80	1.00	0.16
0.20	0.05	0.06	0.01	-0.07	1.00	0.80	1.00	0.11
0.24	0.06	0.37	2.78	0.00	0.53	0.80	1.00	0.03
0.15	0.02	0.09	0.15	-0.03	0.76	0.60	1.20	0.07
0.15	0.04	0.09	0.03	-0.10	0.74	0.60	1.20	0.04
0.33	0.10	0.19	0.05	-0.23	0.83	0.47	1.30	0.18
0.15	0.04	0.09	0.03	-0.07	0.78	0.60	1.20	0.05
0.20	0.05	0.06	0.02	0.07	1.00	0.80	1.00	0.09
0.20	0.05	0.06	0.02	0.03	1.00	0.24	1.00	0.10
0.20	0.01	0.08	0.07	0.01	1.00	0.80	1.00	0.01
0.20	0.05	0.08	0.04	0.06	1.00	0.80	1.00	0.08
0.14	0.02	0.09	0.15	0.07	0.76	0.60	1.20	0.04
0.20	0.03	0.08	0.05	0.20	1.00	0.80	1.00	0.07
0.20	0.03	0.08	0.06	0.03	1.00	0.80	1.00	0.03

0.15	0.02	0.09	0.15	-0.02	0.73	0.60	1.20	0.04
0.20	0.05	0.06	0.02	-0.06	1.00	0.80	1.00	0.12
0.60	0.03	0.12	0.05	-0.03	0.75	0.47	1.30	0.16
0.15	0.03	0.05	0.01	-0.01	0.79	0.06	1.20	0.06
0.15	0.02	0.09	0.15	-0.02	0.74	0.60	1.20	0.05
0.15	0.03	0.05	0.01	-0.02	0.78	0.06	1.20	0.06
0.14	0.02	0.09	0.15	0.00	0.84	0.60	1.20	0.03
0.60	0.03	0.12	0.05	-0.05	0.75	0.47	1.30	0.14
0.15	0.02	0.09	0.15	-0.05	0.75	0.60	1.20	0.05
0.15	0.02	0.09	0.15	-0.04	0.74	0.60	1.20	0.05
0.15	0.04	0.09	0.03	-0.06	0.79	0.60	1.20	0.05
0.30	0.15	0.18	0.03	0.03	1.18	1.00	1.00	0.17
0.20	0.01	0.08	0.07	0.20	1.00	0.80	1.00	0.02
0.20	0.03	0.06	0.03	0.05	1.00	0.80	1.00	0.05
0.20	0.03	0.08	0.05	0.06	1.00	0.80	1.00	0.06
0.20	0.05	0.06	0.02	0.06	1.00	0.24	1.00	0.08
0.20	0.05	0.06	0.01	0.20	1.00	0.80	1.00	0.10
0.20	0.05	0.06	0.02	0.05	1.00	0.24	1.00	0.09
0.15	0.03	0.05	0.01	-0.02	0.78	0.06	1.20	0.06
0.20	0.01	0.08	0.07	0.00	1.00	0.80	1.00	0.05
0.20	0.05	0.08	0.04	0.00	1.00	0.80	1.00	0.11
0.15	0.04	0.09	0.03	0.00	0.76	0.60	1.20	0.09
0.20	0.03	0.06	0.03	0.00	1.00	0.80	1.00	0.08
0.33	0.10	0.19	0.05	-0.10	0.83	0.47	1.30	0.20
0.33	0.10	0.19	0.05	-0.05	0.83	0.47	1.30	0.23
0.20	0.01	0.08	0.07	-0.05	1.00	0.80	1.00	0.10
0.15	0.04	0.09	0.03	-0.07	0.75	0.60	1.20	0.06
0.15	0.02	0.09	0.15	-0.12	0.78	0.60	1.20	0.06
0.15	0.02	0.09	0.15	-0.04	0.75	0.60	1.20	0.06
0.20	0.01	0.08	0.07	0.02	1.00	0.80	1.00	0.02
0.15	0.02	0.09	0.15	0.00	0.76	0.60	1.20	0.01
0.30	0.15	0.21	0.06	0.03	1.18	1.00	1.00	0.22
0.20	0.03	0.08	0.06	0.03	1.00	0.80	1.00	0.03
0.20	0.05	0.08	0.04	0.09	1.00	0.80	1.00	0.10
0.20	0.03	0.06	0.03	0.03	1.00	0.80	1.00	0.06
0.15	0.04	0.09	0.03	0.05	0.75	0.60	1.20	0.06
0.15	0.02	0.09	0.15	-0.01	0.74	0.60	1.20	0.03
0.15	0.03	0.05	0.01	-0.01	0.79	0.06	1.20	0.07

0.20	0.05	0.06	0.01	0.00	1.00	0.80	1.00	0.10
0.15	0.02	0.09	0.15	-0.02	0.76	0.60	1.20	0.06
0.16	0.02	0.09	0.15	0.00	0.75	0.60	1.20	0.03
0.20	0.03	0.08	0.05	0.00	1.00	0.80	1.00	0.09
0.15	0.04	0.09	0.03	-0.01	0.78	0.60	1.20	0.09
0.15	0.02	0.09	0.15	-0.01	0.76	0.60	1.20	0.05
0.14	0.02	0.09	0.15	-0.07	0.80	0.60	1.20	0.07
0.14	0.02	0.09	0.15	-0.18	0.77	0.60	1.20	0.05
0.33	0.10	0.19	0.05	-0.16	0.83	0.47	1.30	0.18
0.33	0.10	0.19	0.05	0.08	0.83	0.47	1.30	0.18
0.13	0.02	0.09	0.15	0.05	0.81	0.60	1.20	0.04
0.20	0.05	0.08	0.04	0.04	1.00	0.80	1.00	0.09
0.20	0.05	0.06	0.02	0.06	1.00	0.24	1.00	0.08
0.20	0.05	0.06	0.02	0.06	1.00	0.24	1.00	0.08
0.20	0.05	0.06	0.02	0.03	1.00	0.80	1.00	0.10
0.20	0.01	0.08	0.07	-0.03	1.00	0.80	1.00	0.09
0.60	0.03	0.12	0.05	0.00	0.75	0.47	1.30	0.15
0.60	0.03	0.12	0.05	-0.33	0.75	0.47	1.30	0.11
0.33	0.10	0.19	0.05	0.00	0.83	0.47	1.30	0.25
0.15	0.04	0.09	0.03	-0.03	0.76	0.60	1.20	0.07
0.15	0.02	0.09	0.15	-0.01	0.76	0.60	1.20	0.03
0.60	0.03	0.12	0.05	0.00	0.84	0.47	1.30	0.04
0.60	0.03	0.12	0.05	0.00	0.84	0.47	1.30	0.04
0.15	0.02	0.09	0.15	-0.17	0.77	0.60	1.20	0.06
0.60	0.03	0.12	0.05	-0.65	0.84	0.47	1.30	0.08
0.15	0.02	0.09	0.15	-0.16	0.79	0.60	1.20	0.06