

# Supplement to “Systemic Risk: Conditional Distortion Risk Measures”

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## Abstract

This online supplementary material provides several numerical illustrations of our main comparison results. For context, notation, and definitions we refer to the paper.

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# 1 Numerical Examples

We provide some numerical examples to illustrate our main findings. Based on our results developed in the paper, the choice of the d.f.'s and distortion functions of  $X$  and  $X'$  can be arbitrary since we only need the relation between  $u_g$  and  $u_{g'}/u_{\tilde{g}}$ . Therefore, we do not specify the explicit d.f.'s of  $X$  and  $X'$  in most of our examples. We shall provide illustrations of our main results both for positive and negative dependence structures, which are represented by the Gumbel copula and the Farlie-Gumbel-Morgenstern (FGM) copula, respectively.

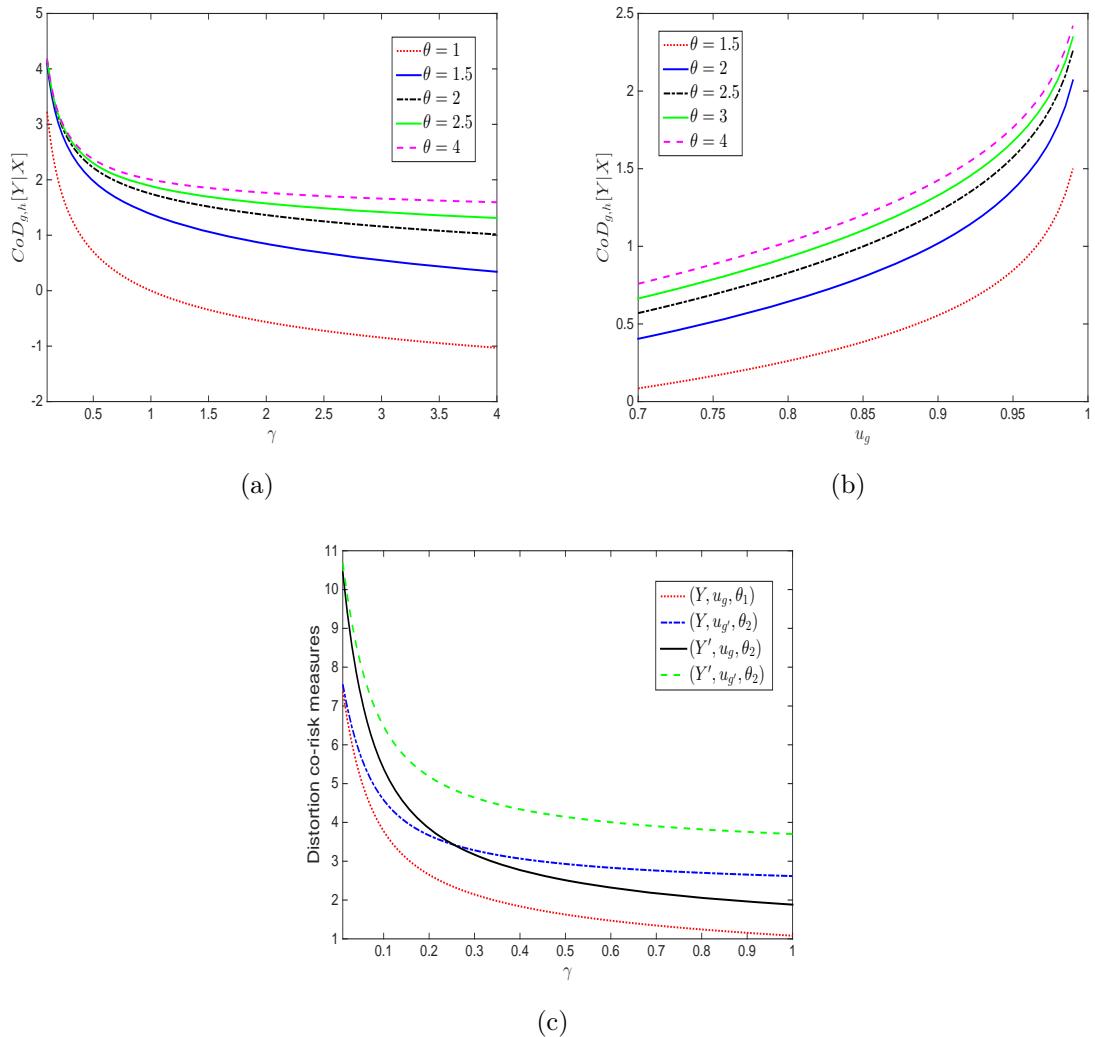


Figure 1: (a) Plot of  $CoD_{g,h}[Y|X]$  on  $\gamma \in [0.1, 4]$  for different values of  $\theta$ . (b) Plot of  $CoD_{g,h}[Y|X]$  on  $u_g \in [0.7, 0.99]$  for different values of  $\theta$ . (c) Plot of CoD-risk measures on  $\gamma \in (0, 1]$  under different settings of d.f., threshold quantile, and dependence parameter.

## 1.1 The Gumbel Copula

The Gumbel copula is defined as

$$C_\theta(u, v) = \exp \left( - \left( (-\log u)^\theta + (-\log v)^\theta \right)^{1/\theta} \right), \quad \theta \geq 1.$$

It corresponds to the independence copula when  $\theta = 1$ , and to the comonotonic copula when  $\theta = +\infty$ . It can be inferred from [Wei and Hu \(2002\)](#) that  $C_\theta \prec C_{\theta'}$  if  $\theta \leq \theta'$ . Besides,  $C_\theta$  is PDS for all  $\theta \geq 1$ . Interested readers are referred to [Joe \(1997\)](#) and [Nelsen \(2007\)](#) for more discussions.

**Example 1.1** (CoD-risk measures). *Assume that  $Y$  has a standard normal d.f. and  $u_g = 0.95$  for some chosen d.f. of  $X$  and distortion function  $g$ . Let  $h(p) = p^\gamma$  for  $\gamma > 0$ . Note that  $h(p)$  is decreasing in  $\gamma$  for any  $p \in [0, 1]$ .*

- (a) *For different values of the dependence parameter  $\theta = 1, 1.5, 2, 2.5, 4$ , we plot the values of  $\text{CoD}_{g,h}[Y|X]$  for  $\gamma > 0$  in Figure 1(a). It is readily apparent that the CoD-risk measure decreases as the distortion function of  $Y$  gets smaller (i.e.,  $\gamma$  gets larger) for fixed dependence parameter  $\theta$ , and it increases when the positive dependence gets stronger (i.e.,  $\theta$  gets larger). This illustrates the result of Theorem 4.1.*
- (b) *For different values of the dependence parameter  $\theta = 1.5, 2, 2.5, 3, 4$ , we plot the values of  $\text{CoD}_{g,h}[Y|X]$  as  $u_g$  varies from 0.7 to 0.99 in Figure 1(b), from which we observe that the CoD-risk measure increases as the threshold quantile  $u_g$  gets larger for fixed dependence parameter  $\theta$ , and it increases when the positive dependence gets stronger (i.e.,  $\theta$  gets larger). Therefore, the theoretical finding in Theorem 4.4(i) is verified.*
- (c) *Consider  $Y \sim N(0, 1)$  and  $Y' \sim N(0, 2)$  such that  $Y \leq_{\text{icx}} Y'$  but  $Y \not\leq_{\text{st}} Y'$ . Assume that  $\theta_1 = 2$ ,  $\theta_2 = 4$ ,  $u_g = 0.8$ , and  $u_{g'} = 0.99$ . Figure 1(c) gives the plots of  $\text{CoD}_{g,h}[Y|X]$ ,  $\text{CoD}_{g',h}[Y|X']$ ,  $\text{CoD}_{g,h}[Y'|X]$ , and  $\text{CoD}_{g',h}[Y'|X']$  for different values of  $\gamma \in (0, 1]$ , which implies that  $h(p)$  is increasing and concave on  $p \in [0, 1]$ . It is readily apparent that these four types of CoD-risk measures become smaller as  $\gamma$  increases, i.e., as the distortion function becomes smaller. Moreover, for any fixed  $\gamma \in (0, 1]$ , we have*

$$\text{CoD}_{g,h}[Y|X] \leq \text{CoD}_{g',h}[Y|X'] \leq \text{CoD}_{g',h}[Y'|X'],$$

$$\text{CoD}_{g,h}[Y|X] \leq \text{CoD}_{g,h}[Y'|X] \leq \text{CoD}_{g',h}[Y'|X'],$$

while  $\text{CoD}_{g',h}[Y|X']$  and  $\text{CoD}_{g,h}[Y'|X]$  cannot be compared. These observations validate the results of Theorem 4.11(i).

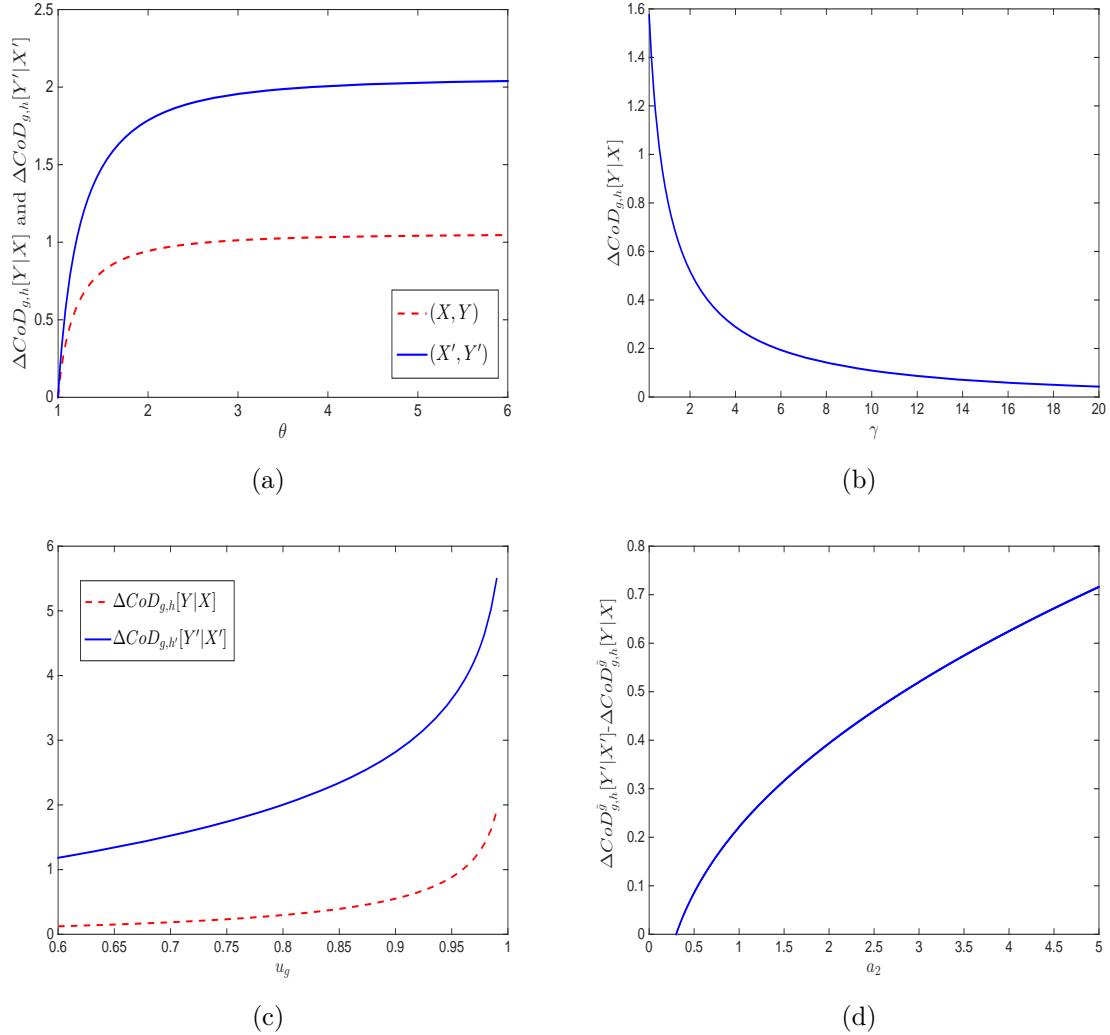


Figure 2: (a) Plot of  $\Delta CoD_{g,h}[Y|X]$  and  $\Delta CoD_{g,h}[Y'|X']$  on  $\theta \geq 1$ . (b) Plot of  $\Delta CoD_{g,h}[Y|X]$  on  $\gamma > 0$ . (c) Plot of  $\Delta CoD_{g,h}[Y|X]$  and  $\Delta CoD_{g,h'}[Y'|X']$  on  $u_g \in [0.6, 0.99]$ . (d) Plot of  $\Delta CoD_{g,h}^{\tilde{g}}[Y'|X'] - \Delta CoD_{g,h}^{\tilde{g}}[Y|X]$  for different values of the shape parameter  $a_2 \geq a_1$ .

The next example supports our comparison results for the distortion risk contribution measures.

**Example 1.2** (Distortion risk contribution measures). *In this example, we assume that the distortion functions applied to  $Y$  and  $Y'$  are of the form of a power function.*

- (a) Suppose that  $Y \sim \Gamma(a_1, b_1)$  and  $Y' \sim \Gamma(a_2, b_2)$  with  $(a_1, b_1) = (0.3, 1)$  and  $(a_2, b_2) = (2, 1)$ . Thus, it holds that  $Y \leq_{\text{disp}} Y'$ . Let  $u_g = 0.8$  and  $h(p) = p^{0.4}$ , for  $p \in [0, 1]$ . Figure 2(a) displays the plots of  $\Delta\text{CoD}_{g,h}[Y|X]$  and  $\Delta\text{CoD}_{g,h}[Y'|X']$  on  $\theta \geq 1$ , from which one can observe that  $\Delta\text{CoD}_{g,h}[Y|X] \leq \Delta\text{CoD}_{g,h}[Y'|X']$  for any fixed  $\theta$ , and both of them are increasing with respect to ‘ $\prec$ ’. This supports the result of Theorem 5.2(i).
- (b) Let  $Y \sim \Gamma(0.2, 1)$ ,  $u_g = 0.9$ , and  $\theta = 2$ . It is clear that  $Y$  is DFR. The value of  $\Delta\text{CoD}_{g,h}[Y|X]$  is plotted in Figure 2(b) for different distortion functions applied to  $Y$ . It is straightforward to observe that  $\Delta\text{CoD}_{g,h}[Y|X]$  is decreasing with respect to  $\gamma$ , which verifies Theorem 5.5(i).
- (c) Let  $h(p) = p^{\gamma_1}$ ,  $h'(p) = p^{\gamma_2}$ ,  $C$  with parameter  $\theta_1$  and  $C'$  with parameter  $\theta_2$ . Set  $\gamma_1 = 3$ ,  $\gamma_2 = 2$ ,  $\theta_1 = 2$ ,  $\theta_2 = 3$ ,  $Y \sim \Gamma(0.2, 1)$ , and  $Y' \sim \Gamma(2, 1)$ . Figure 2(c) plots  $\Delta\text{CoD}_{g,h}[Y|X]$  and  $\Delta\text{CoD}_{g,h'}[Y'|X']$  on  $u_g \in [0.6, 0.99]$ . We observe that both  $\Delta\text{CoD}_{g,h}[Y|X]$  and  $\Delta\text{CoD}_{g,h'}[Y'|X']$  are increasing with respect to  $u_g$ , and  $\Delta\text{CoD}_{g,h}[Y|X] \leq \Delta\text{CoD}_{g,h'}[Y'|X']$  for any fixed  $u_g$ , which validates the result of Theorem 5.8.
- (d) Assume that  $u_g = 0.9$ ,  $u_{\tilde{g}} = 0.8$ ,  $h(p) = p^2$ ,  $\theta = 2$ ,  $Y \sim \Gamma(a_1, 1)$ , and  $Y' \sim \Gamma(a_2, 1)$  with  $a_2 > 0$ . The difference function between  $\Delta\text{CoD}_{g,h}^{\tilde{g}}[Y'|X']$  and  $\Delta\text{CoD}_{g,h}^{\tilde{g}}[Y|X]$  is plotted in Figure 2(d), which is always negative for all  $a_2 \geq a_1 = 0.3$ . Thus, the result of Theorem 5.9 is validated.

Next, we present an example to illustrate the condition in Theorem 5.13.

**Example 1.3.** Assume that  $h(p) = 1 - (1 - p)^\gamma$  for  $\gamma > 1$ . Let  $C$  be the Gumbel copula with dependence parameter  $\theta > 1$ . It is easy to verify that  $h(p)$  is concave and  $\bar{h}(p) = p^\gamma$ . Observe that

$$\Psi(t) := \bar{h}(A(\bar{h}^{-1}(t))) = \left[ \frac{t^{\frac{1}{\gamma}} - C(u_g, t^{\frac{1}{\gamma}})}{1 - u_g} \right]^\gamma.$$

- (a) Set  $u_g = 0.9$  and  $\gamma = 1.1$ . Figure 3(a) plots  $\Psi(t)$  on  $t \in [0, 1]$  under different values of  $\theta = 1.2, 1.8, 2.5, 3, 5$ , which indicates the convexity of  $\Psi(t)$ .
- (b) Set  $u_g = 0.9$  and  $\theta = 1.5$ . Figure 3(b) plots  $\Psi(t)$  on  $t \in [0, 1]$  under different values of  $\gamma = 1.2, 2, 3, 4, 5$ , from which one can observe the convexity of  $\Psi(t)$ .

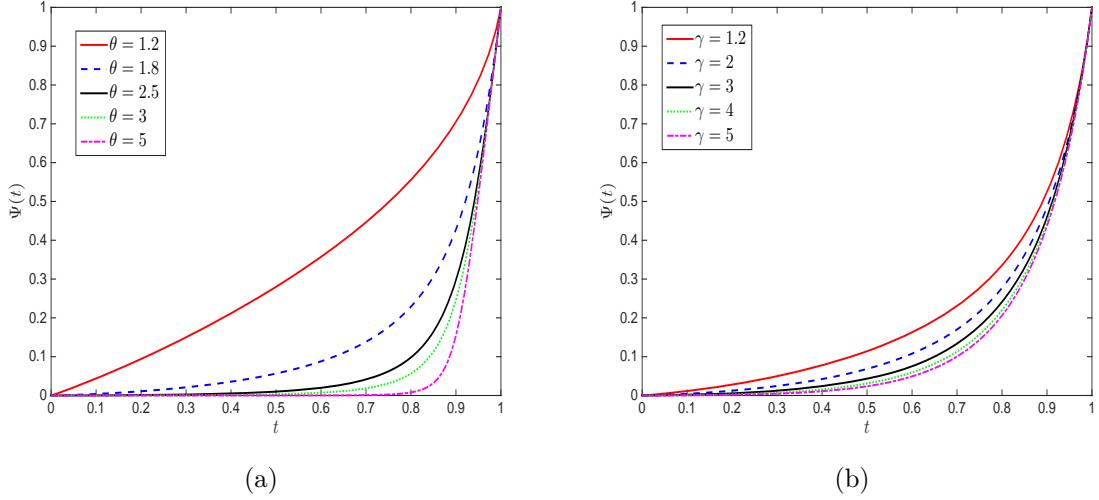


Figure 3: (a) Plot of  $\Psi(t)$  on  $t \in [0, 1]$  for different values of  $\theta$ . (b) Plot of  $\Psi(t)$  on  $t \in [0, 1]$  for different values of  $\gamma$ .

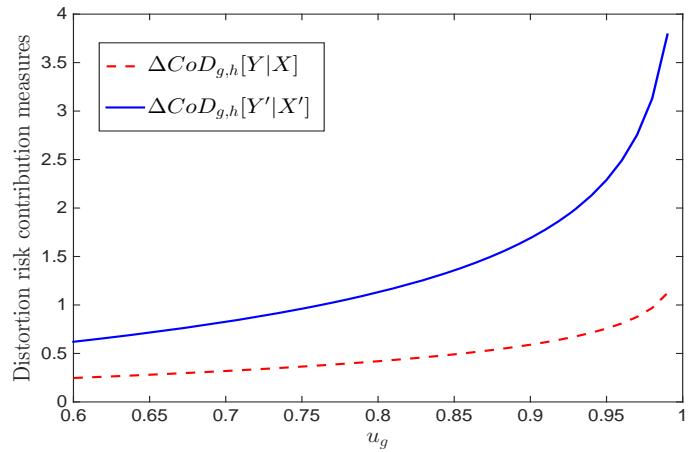


Figure 4: Plot of  $\Delta\text{CoD}_{g,h}[Y|X]$  and  $\Delta\text{CoD}_{g,h}[Y'|X']$  for  $u_g \in [0.6, 0.99]$ .

The following example illustrates Theorem 5.13.

**Example 1.4.** Assume that  $h(p) = 1 - (1 - p)^2$ ,  $\theta = 1.5$ ,  $Y \sim W(1, 2)$ , and  $Y \sim W(1, 1)$ . Clearly, it holds that  $Y \leq_{\text{ew}} Y'$  but  $Y \not\leq_{\text{disp}} Y'$  nor  $Y \not\geq_{\text{disp}} Y'$  (see Example 24 in Sordo et al., 2018). As displayed in Figure 4,  $\Delta\text{CoD}_{g,h}[Y|X] \leq \Delta\text{CoD}_{g,h}[Y'|X']$  for  $u_g \in [0.6, 0.99]$ , which shows the effectiveness of Theorem 5.13.

Next, we present a numerical example to show the effectiveness of Theorem 6.1.

**Example 1.5.** Let  $C$  be the Gumbel copula with dependence parameter  $\theta = 2$ . Assume that  $g(t) = t^{0.3}$ ,  $X \sim \Gamma(0.5, 1)$ ,  $Y \sim \Gamma(1.5, 1)$ , and  $h(p) = p^{\gamma_2}$  for  $\gamma_2 > 0$ . It is easy to verify that  $X \leq_{\text{st}} Y$  and  $X \leq_{\text{disp}} Y$ . Moreover, one can calculate that  $u_g^X = 0.9714 > u_g^Y = 0.9599$ . Figure 5(a) displays  $\text{CoD}_{g,h}[Y|X]$  and  $\text{CoD}_{g,h}[X|Y]$  for  $\gamma_2 > 0$ , and Figure 5(b) plots  $\Delta\text{CoD}_{g,h}[Y|X]$  and  $\Delta\text{CoD}_{g,h}[X|Y]$  for  $\gamma_2 > 0$ . Obviously,  $\text{CoD}_{g,h}[Y|X] \geq \text{CoD}_{g,h}[X|Y]$  and  $\Delta\text{CoD}_{g,h}[Y|X] \geq \Delta\text{CoD}_{g,h}[X|Y]$  for  $\gamma_2 > 0$ . Therefore, the results of Theorem 6.1(i) and Theorem 6.1(v) are supported.

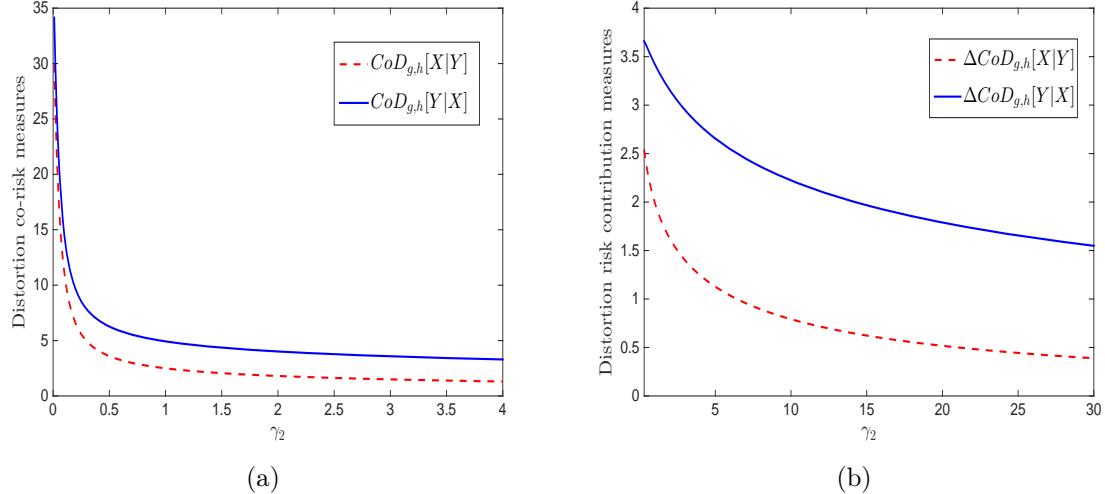


Figure 5: (a) Plot of  $\text{CoD}_{g,h}[Y|X]$  and  $\text{CoD}_{g,h}[X|Y]$  for  $\gamma_2 > 0$ . (b) Plot of  $\Delta\text{CoD}_{g,h}[Y|X]$  and  $\Delta\text{CoD}_{g,h}[X|Y]$  for  $\gamma_2 > 0$ .

## 1.2 The Farlie-Gumbel-Morgenstern Copula

The Farlie-Gumbel-Morgenstern (FGM) copula is defined as

$$C_\alpha(u, v) = uv [1 + \alpha(1 - u)(1 - v)], \quad -1 \leq \alpha \leq 1.$$

If  $\theta = 0$ , then  $C_\theta$  reduces to the independence copula. Furthermore,  $C_\alpha(u, v)$  is RR<sub>2</sub> [TP<sub>2</sub>] for  $\alpha \in [-1, 0)$  [ $\alpha \in [0, 1]$ ] and  $\alpha_1 \leq \alpha_2$  implies that  $C_{\alpha_1} \prec C_{\alpha_2}$ . For more details on its properties, we refer to Joe (1997) and Nelsen (2007).

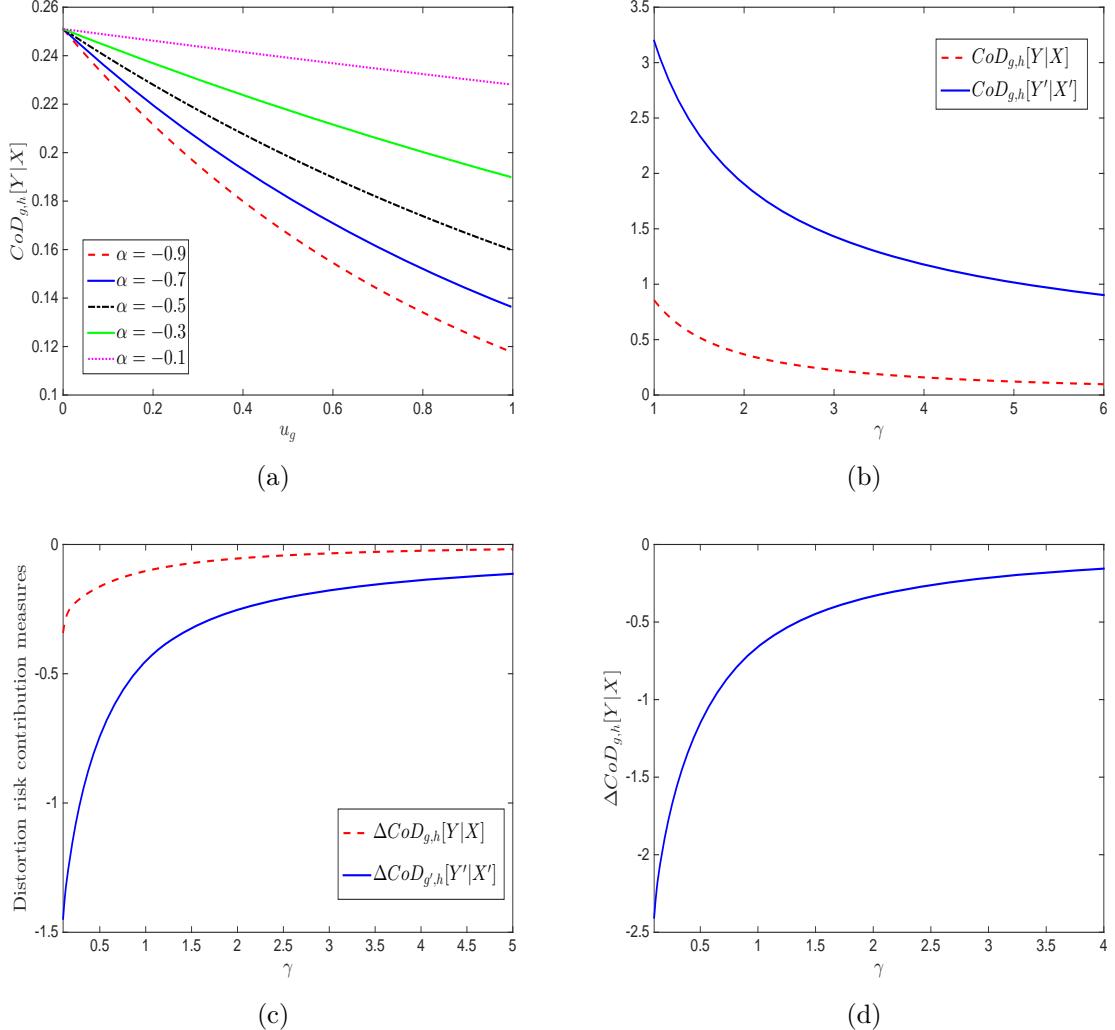


Figure 6: (a) Plot of  $CoD_{g,h}[Y|X]$  for  $u_g \in [0, 1]$ . (b) Plot of  $\Delta CoD_{g,h}[Y|X]$  and  $\Delta CoD_{g,h}[Y'|X']$  for  $\gamma \geq 1$ . (c) Plot of  $\Delta CoD_{g,h}[Y|X]$  and  $\Delta CoD_{g',h}[Y'|X']$  for  $\gamma > 0$ . (d) Plot of  $\Delta CoD_{g,h}[Y|X]$  for  $\gamma > 0$ .

The following examples show the effectiveness of Theorem 4.4(ii), Theorem 4.7(ii), Theorem 5.3(ii), Theorem 5.5(ii), and Theorem 6.1 under the negative dependence characterized by the FGM copula.

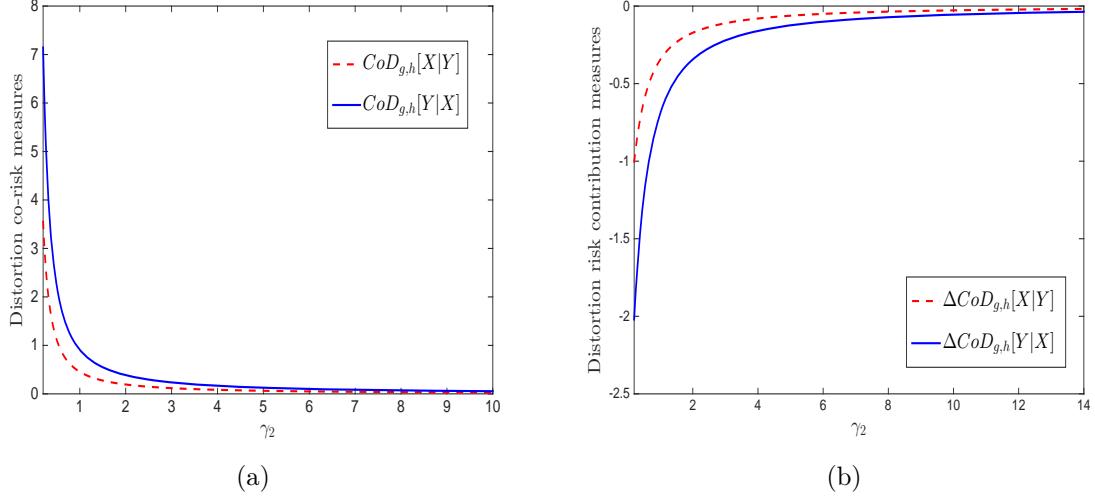


Figure 7: (a) Plot of  $CoD_{g,h}[Y|X]$  and  $CoD_{g,h}[X|Y]$  for  $\gamma_2 > 0$ . (b) Plot of  $\Delta CoD_{g,h}[Y|X]$  and  $\Delta CoD_{g,h}[X|Y]$  for  $\gamma_2 > 0$ .

**Example 1.6.** (a) Set  $Y \sim \Gamma(0.8, 2)$  and  $h(p) = p^5$  for  $p \in [0, 1]$ . Figure 6(a) displays  $CoD_{g,h}[Y|X]$  on  $u_g \in [0, 1]$  for different values of the dependence parameter  $\alpha = -0.9, -0.7, -0.5, -0.3, -0.1$ . One readily observes that  $CoD_{g,h}[Y|X]$  is decreasing with respect to  $u_g$  for any fixed  $\alpha$ , while it is increasing in  $\alpha$  for any fixed  $u_g$ . This agrees with the result of Theorem 4.4(ii).

- (b) Set  $Y \sim \Gamma(0.8, 2)$ ,  $Y' \sim \Gamma(1.8, 2)$ ,  $u_g = 0.95$ ,  $\alpha_1 = -0.9$ , and  $\alpha_2 = -0.3$ . Let  $h(p) = p^\gamma$  for  $\gamma \geq 1$  and  $p \in [0, 1]$ , which means that  $h$  is increasing and convex. The values of  $\Delta CoD_{g,h}[Y|X]$  and  $\Delta CoD_{g,h}[Y'|X']$  are plotted in Figure 6(b) for  $\gamma \geq 1$ , from which it is clear that  $\Delta CoD_{g,h}[Y|X] \leq \Delta CoD_{g,h}[Y'|X']$  for  $\gamma \geq 1$ . Thus, the result of Theorem 4.7(ii) is validated.
- (c) Suppose that  $Y \sim \Gamma(0.6, 1)$ ,  $Y' \sim \Gamma(1.2, 1)$ ,  $u_g = 0.95$ ,  $u_{g'} = 0.9$ ,  $\alpha = -0.3$ , and  $\alpha' = -0.9$ . It is plain that  $Y \leq_{\text{disp}} Y'$ ,  $u_g \geq u_{g'}$ , and  $C' \prec C$ . As observed from Figure 6(c),  $\Delta CoD_{g,h}[Y|X] \geq \Delta CoD_{g',h}[Y'|X']$  for  $\gamma > 0$ , which illustrates Theorem 5.3(ii).
- (d) Set  $\alpha = -0.8$ ,  $Y \sim \Gamma(0.8, 2)$ , and  $u_g = 0.95$ . Thus,  $Y$  is DFR. Let  $h(p) = p^\gamma$  for  $\gamma > 0$ . As shown in Figure 6(d), the value of  $\Delta CoD_{g,h}[Y|X]$  is increasing with respect to  $\gamma > 0$ , which validates the theoretical finding of Theorem 5.5(ii).

Finally, a numerical example is provided to show the effectiveness of Theorem 6.1 under the case of negative dependence.

**Example 1.7.** Let  $C$  be the FGM copula with dependence parameter  $\alpha = -0.8$ . Assume that  $g(p) = p^{0.2}$ ,  $X \sim \Gamma(0.8, 1)$ ,  $Y \sim \Gamma(0.8, 2)$ , and  $h(p) = p^{\gamma_2}$  for  $\gamma_2 > 0$ . It is easy to verify that  $X \leq_{\text{hr}} Y$  and thus  $X \leq_{\text{st}} [ \text{disp} ] Y$  since both  $X$  and  $Y$  are DFR. Moreover, one can calculate that  $u_g^X = u_g^Y = 0.9937$ . Figure 7(a) displays  $\text{CoD}_{g,h}[Y|X]$  and  $\text{CoD}_{g,h}[X|Y]$  for  $\gamma_2 > 0$ , and Figure 7(b) plots  $\Delta\text{CoD}_{g,h}[Y|X]$  and  $\Delta\text{CoD}_{g,h}[X|Y]$  for  $\gamma_2 > 0$ . Note that  $\text{CoD}_{g,h}[Y|X] \geq \text{CoD}_{g,h}[X|Y]$  while  $\Delta\text{CoD}_{g,h}[Y|X] \leq \Delta\text{CoD}_{g,h}[X|Y]$  for all  $\gamma_2 > 0$ , and thus the results of Theorems 6.1(ii) and 6.1(iv) are validated.

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