# A SIMPLE TWO-PLAYER TWO-SIDED GAMES OF DECEPTION 

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#### Abstract

Let $X_{1}, X_{2}, Y_{1}, Y_{2}$ are i.i.d. r.v.s obeying the same probability destribution. Player I [II] looks privately $\left(X_{1}, X_{2}\right)=\left(x_{1}, x_{2}\right),\left[\left(Y_{1}, Y_{2}\right)=\left(y_{1}, y_{2}\right)\right]$. They choose a single common number $\theta$, and player II opens the nearest number to $\theta$ among $y_{1}$ and $y_{2}$ and covers the other number. If II's opened number is $>(<) \theta$, then I gets as his payoff, the opened (covered) number. The problem is to determine $\theta$ under which the expected payoff $M_{1}(\theta)$, I can get, is maximized. The maximization problem for II is quite similar as for I. Because of symmetry between the two players, our problem essentially reduces to computation of $M_{1}(\theta)$ and finding the $\theta$ which maximizes $M_{1}(\theta)$. This game is solved for (1) uniform distribution on [0, 1], (2) exponential distribution on $[0, \infty)$, (3) normal distribution on $(-\infty, \infty)$, and (4) some other distributions around them.


1 Two-player One-sided Games of Deception. Two numbers $x_{1}$ and $x_{2}$ are chosen from $[0,1]$ by means of independent bivariate uniform distribution on $[0,1]^{2}$. Player I now looks at the numbers privately and chooses one of the two and opens it to Player II, and the other number is covered. Player II then accepts either one of the opened number or the covered number, and receives from player I the number he accepted. Player I (II) aims to minimize (maximize) the expected payoff to II.

In Baston and Bostock (Ref.[1]) it is proven that the strategies;
$\sigma^{*}$ : Choose the nearest number to $\frac{1}{2}$ among $x_{1}$ and $x_{2}$, and open it. The other number is covered.
for I and
$\tau^{*}$ : Accept the opened (covered) number if it is $>(<) \frac{1}{2}$,
for II, constitute an optimal strategy-pair, and the value of the game is $7 / 12$.
By Sakaguchi (Ref.[5]) it is proven that, if $x_{1}$ and $x_{2}$ are independent bivariate standard normal distribution in $(-\infty, \infty)^{2}$, then the above strategy-pair $\sigma^{*}$ and $\tau^{*}$ with $\frac{1}{2}$ replaced by 0 , is optimal, and the value of the game is $\frac{2-\sqrt{2}}{\sqrt{2 \pi}} \approx 0.2337$.

In the present paper, we make an approach to consider the two-sided game, where each player aims to maximize his expected payoff he obtains from his opponent.

2 Two-player Two-sided Games of Deception. Let $X_{1}, X_{2}, Y_{1}, Y_{2}$ are i.i.d. r.v.s with an identical p.d.f. Player I observes $\left(X_{1}, X_{2}\right)$ and chooses his decision number $\theta_{1} \in$ $[0,1]$. Player II observes $\left(Y_{1}, Y_{2}\right)$ and chooses his decision number $\theta_{2} \in[0,1]$. Each player's choice of his decision number is made independently of the opponent's choice.

Player I chooses the nearest number to $\theta_{1}$ among $x_{1}$ and $x_{2}$ and open it and the other number is covered. Player II chooses the nearest number to $\theta_{2}$ among $y_{1}$ and $y_{2}$ and opens it and the other number is covered. If II's opened number is $>(<) \theta_{1}$, then I gets II's opened (covered) number. If I's opened number is $>(<) \theta_{2}$, then II gets I's opened (covered)

[^0]number. For the sake of symmetry it should be $\theta_{1}=\theta_{2}(=\theta$, say $)$. Both players want to choose the optimal $\theta$ which maximizes the common expected payoff, they can get.

An example of non-simple two-sided games of deception is mentioned in Remark 3 of Section 6.

3a Uniform Destribution in $[0,1]$. We divide the plane $[0,1]^{2}$ by the four "quadrant" s by the two axis $y_{1}=\theta$ and $y_{2}=\theta$, and denote them $Q^{(1)}, Q^{(2)}, Q^{(3)}$ and $Q^{(4)}$, in the clock-wise order. We use this convention throughout this paper.

We consider the two case $0 \leq \theta<\frac{1}{2}$ and $\frac{1}{2} \leq \theta \leq 1$. First let $0 \leq \theta<\frac{1}{2}$, Player I can get his payoff shown as in Figure 1.


Figure 1. Case $0 \leq \theta<\frac{1}{2}$


Figure 2. Case $\frac{1}{2} \leq \theta \leq 1$

The proof is as follows. In the upper-left of $Q^{(1)}, y_{1}+y_{2}>2 \theta$ and $\theta<y_{1}<y_{2}$ hold, and hence II opens $y_{1}$, and I gets $y_{1}$. In the lower-right of $Q^{(1)}, y_{1}+y_{2}>2 \theta$ and $y_{1}>y_{2}>\theta$, and hence II opens $y_{2}$ and I gets $y_{2}$. In the upper-right of $Q^{(2)}, y_{1}+y_{2}>2 \theta$ and $y_{2}<y_{1} \wedge \theta$, and so II opens $y_{2}$, and I gets the covered $y_{1}$. In the lower-left of $Q^{(2)}, y_{1}+y_{2}<2 \theta$ and $y_{1}>y_{2} \vee \theta$, and so II opens $y_{1}$ and I gets $y_{1}$. In $Q^{(3)}$ and $Q^{(4)}$ similar arguments can be made and the result is as shown in Figure 1.

I's total expected reward is

$$
\begin{equation*}
2\left[\int_{\theta}^{1} d y_{2} \int_{\theta}^{y_{2}} y_{1} d y_{1}+\int_{0}^{\theta} d y_{2} \int_{\theta}^{1} y_{1} d y_{1}+\int_{0}^{\theta} d y_{2} \int_{0}^{y_{2}} y_{1} d y_{1}\right] \tag{3.1}
\end{equation*}
$$

The reason of why (3.1) is 2 times of $[\cdots]$ is that " $y_{1}$ domain" and " $y_{2}$ domain" are located symmetric about the straight line $y_{1}=y_{2}$, and $y_{1}$ and $y_{2}$ are i.i.d. distributed.
Here, the first (second, third) term gives the reward from $Q^{(1)}\left(Q^{(2)}, Q^{(3)}\right)$. The sum of these three is equal to

$$
\begin{equation*}
2\left[\left(\frac{1}{6}-\frac{1}{2} \theta^{2}+\frac{1}{3} \theta^{3}\right)+\frac{1}{2}\left(\theta-\theta^{3}\right)+\frac{1}{6} \theta^{3}\right]=\frac{1}{3}+\theta-\theta^{2} \tag{3.2}
\end{equation*}
$$

which is concave, increasing with values $\frac{1}{3}$ at $\theta=0$ and $\frac{7}{12}$ at $\theta=\frac{1}{2}$.
Now secondly, let $\frac{1}{2} \leq \theta \leq 1$. We repeat the same arguments as in the case $0 \leq \theta<\frac{1}{2}$ (see Figure 2) and we obtain the same result $\frac{1}{3}+\theta-\theta^{2}$, which is concave, decreasing with values $\frac{7}{12}$ at $\theta=\frac{1}{2}$ and $\frac{1}{3}$ at $\theta=1$.

Thus we can state

Theorem 1 In Case $3 a$, the optimal choice is $\theta^{*}=\frac{1}{2}$ and the common OPR (optimal reward) is $7 / 12$.

3b Triangular Distribution in $[0,1]$. The p.d.f. is $f(x)=2 x, x \in[0,1]$. Mean value is $2 / 3$. Figures 1 and 2 remain unchanged as in section 3a. We explain in the following the values in $Q^{(3)}$ and $Q^{(4)}$. In the upper-left of $Q^{(3)}$ in Figure 1, we have $y_{1}+y_{2}<2 \theta$ and $y_{1}<y_{2}<\theta$ and so II opens $y_{2}$ and I gets the covered number $y_{1}$. In the lower-right of $Q^{(3)}, y_{1}+y_{2}<2 \theta$ and $\theta>y_{1}>y_{2}$, and so II opens $y_{1}$ and I gets the covered number $y_{2}$. In the upper-right of $Q^{(4)}$ we have $y_{1}+y_{2}>2 \theta$ and $y_{1}<y_{2} \wedge \theta$, and so II opens $y_{1}$ and I gets the covered number $y_{2}$. In the upper-left of $Q^{(4)}$ we have $y_{1}+y_{2}<2 \theta$ and $y_{2}>y_{1} \vee \theta$, and so II opens $y_{2}$ and I gets $y_{2}$. The result is as shown in Figure 1.

I's total expected reward is

$$
\begin{align*}
& 8\left[\int_{\theta}^{1} y_{2} d y_{2} \int_{0}^{y_{2}} y_{1}^{2} d y_{1}+\int_{0}^{\theta} y_{2} d y_{2} \int_{\theta}^{1} y_{1}^{2} d y_{1}+\int_{0}^{\theta} y_{2} d y_{2} \int_{0}^{y_{2}} y_{1}^{2} d y_{1}\right]  \tag{3.3}\\
& \quad=8\left[\left(\frac{1}{15}-\frac{1}{6} \theta^{3}+\frac{1}{10} \theta^{5}\right)+\frac{1}{6}\left(\theta^{2}-\theta^{3}\right)+\frac{1}{15} \theta^{5}\right]=8\left(\frac{1}{15}+\frac{1}{6} \theta^{2}-\frac{1}{6} \theta^{3}\right)
\end{align*}
$$

which is increasing and convex-concave (with the point of inflexion $\theta=1 / 3$ ), with values $\frac{8}{15} \approx 0.5333$ at $\theta=0, \frac{256}{405} \approx 0.6321$ at $\theta=\frac{1}{3}$, and 0.7 at $\theta=\frac{1}{2}$.

As for Figure 2, i.e. the case $\frac{1}{2} \leq \theta \leq 1$, computation is made similarly, and we find that I's total expected reward remains unchanged and is (3.3) again. However, this (3.3) for $\frac{1}{2} \leq \theta \leq 1$ is unimodal, concave with the maximal value $\frac{296}{405} \approx 0.7309$ at $\theta=\frac{2}{3}$. I's total expected reward as a function of $\theta \in[0,1]$ is shown by Figure 3 .


Figure 3. The function of $\theta \in[0,1]$
Theorem 2 In Case $3 b$, the optimal choice is $\theta^{*}=\frac{2}{3}$ and the common OPR is $\frac{296}{405} \approx$ 0.7309 .

Moreover, we can show in the same way that
Theorem $2^{\prime}$ In Case $3 b^{\prime}$, where $f(x)=2 \bar{x}$, in $[0,1]$, the optimal choice is $\theta^{*}=\frac{1}{3}$, and the common OPR is $\frac{161}{405} \approx 0.39753$.

It is interesting that $\frac{296}{405}-\frac{161}{405}=\frac{1}{3}$.
4a Exponential Distribution in $[0, \infty)$. The p.d.f. is $f(x)=e^{-x}, x \in[0, \infty)$. Mean value is 1. The analysis goes just as before and we obtain Figure 4, which shows the reward I can get.


Figure 4. The reward player II can get.
The computation as made in (2.1) is 2 times of

$$
\begin{align*}
& \int_{\theta}^{\infty} e^{-y_{2}} d y_{2} \int_{\theta}^{y_{2}} y_{1} e^{-y_{1}} d y_{1}+\int_{0}^{\theta} e^{-y_{2}} d y_{2} \int_{\theta}^{\infty} y_{1} e^{-y_{1}} d y_{1}  \tag{4.1}\\
&+\int_{0}^{\theta} e^{-y_{2}} d y_{2} \int_{0}^{y_{2}} y_{1} e^{-y_{1}} d y_{1}
\end{align*}
$$

where the first integral is

$$
\int_{\theta}^{\infty}-\left(1+y_{2}\right) e^{-2 y_{2}} d y_{2}+(1+\theta) e^{-2 \theta}
$$

the second integral is $(1+\theta)\left(e^{-\theta}-e^{-2 \theta}\right)$, and third integral is

$$
\left(1-e^{-\theta}\right)+\int_{0}^{\theta}\left\{-\left(1+y_{2}\right) e^{-2 y_{2}}\right\} d y_{2}
$$

Hence the sum is

$$
\begin{aligned}
\int_{0}^{\infty}-\left(1+y_{2}\right) e^{-2 y_{2}} d y_{2} & +(1+\theta) e^{-2 \theta}+(1+\theta)\left(e^{-\theta}-e^{-2 \theta}\right)+\left(1-e^{-\theta}\right) \\
& =-\frac{3}{4}+\left(1+\theta e^{-\theta}\right)=\frac{1}{4}+\theta e^{-\theta}
\end{aligned}
$$

which is maximized at $\theta=1$. Thus we have
Theorem 3 In Case $4 a$, the optimal choice is $\theta^{*}=1$, and the common OPR is $\frac{1}{2}+2 e^{-1} \approx$ 1.23576 .

4b Another Exponential Distribution in $[0, \infty)$. The p.d.f. is $f(x)=x e^{-x}, x \in$ $[0, \infty)$. Mean value is 2 . The analysis goes just as before and we obtain Figure 4 again. The computation as in (4.1) is now 2 times of

$$
\begin{align*}
& \int_{\theta}^{\infty} y_{2} e^{-y_{2}} d y_{2} \int_{\theta}^{y_{2}} y_{1}^{2} e^{-y_{1}} d y_{1}+\int_{0}^{\theta} y_{2} e^{-y_{2}} d y_{2} \int_{\theta}^{\infty} y_{1}^{2} e^{-y_{1}} d y_{1}  \tag{4.2}\\
&+\int_{0}^{\theta} y_{2} e^{-y_{2}} d y_{2} \int_{0}^{y_{2}} y_{1}^{2} e^{-y_{1}} d y_{1}
\end{align*}
$$

where the first integral is

$$
\begin{aligned}
& \int_{\theta}^{\infty} y_{2} e^{-y_{2}} d y_{2}\left[\left(-y^{2}-2 y-2\right) e^{-y}\right]_{\theta}^{y_{2}} \\
&=\int_{\theta}^{\infty}-\left(y_{2}^{3}+2 y_{2}^{2}+2 y_{2}\right) e^{-2 y_{2}} d y_{2}+(1+\theta)\left(\theta^{2}+2 \theta+2\right) e^{-2 \theta}
\end{aligned}
$$

the second integral is

$$
\left\{1-(1+\theta) e^{-\theta}\right\}\left(\theta^{2}+2 \theta+2\right) e^{-\theta}=\left(\theta^{2}+2 \theta+2\right)\left\{e^{-\theta}-(1+\theta) e^{-2 \theta}\right\}
$$

and the third integral is

$$
\begin{aligned}
& \int_{0}^{\theta} y_{2} e^{-y_{2}} d y_{2}\left[\left(-y^{2}-2 y-2\right) e^{-y}\right]_{0}^{y_{2}} \\
&=\int_{0}^{\infty}-\left(y_{2}^{3}+2 y_{2}^{2}+2 y_{2}\right) e^{-2 y_{2}} d y_{2}+2\left\{1-(1+\theta) e^{-\theta}\right\}
\end{aligned}
$$

Hence the sum is

$$
\begin{aligned}
-\int_{0}^{\infty}\left(y_{2}^{3}+2 y_{2}^{2}+y_{2}\right) & e^{-2 y_{2}} d y_{2}+\left(\theta^{2}+2 \theta+2\right) e^{-\theta}+2\left\{1-(1+\theta) e^{-\theta}\right\} \\
& =-1+\left[2+\left\{\theta^{2}+2 \theta+2-2(1+\theta)\right\} e^{-\theta}\right]=1+\theta^{2} e^{-\theta}
\end{aligned}
$$

which is maximized at $\theta=2$.
For the computations of these integrals we used the formulas

$$
\begin{gather*}
\int t e^{-t} d t=-(1+t) e^{-t}, \quad \int t^{2} e^{-t} d t=-\left(t^{2}+2 t+2\right) e^{-t}  \tag{4.3}\\
\int t e^{-2 t} d t=-\frac{1}{4}(1+t) e^{-t} \text { and } \int t^{2} e^{-2 t} d t=-\frac{1}{8}\left(t^{2}+2 t+2\right) e^{-t}
\end{gather*}
$$

Thus we obtain
Theorem 4 In Case 4 , the optimal choice is $\theta^{*}=2$, and the common OPR is $2(1+$ $\left.4 e^{-2}\right) \approx 3.0827$.

5a Normal Distribution in $(-\infty, \infty)$. We consider the case $\theta \geq 0$. The case $\theta<0$ is not needed to consider, because of symmetry of the p.d.f.. The reward player II can get is shown, just as in the preceding distribution, by Figure 5.


Figure 5 . Case $\theta \geq 0$.

I's total expected reward is 2 times of

$$
\begin{align*}
& \int_{\theta}^{\infty} \phi\left(y_{2}\right) d y_{2} \int_{\theta}^{y_{2}} y_{1} \phi\left(y_{1}\right) d y_{1}+\Phi(\theta) \int_{\theta}^{\infty} y_{1} \phi\left(y_{1}\right) d y_{1}  \tag{5.1}\\
&+\int_{-\infty}^{\theta} y_{1} \phi\left(y_{1}\right)\left(\Phi(\theta)-\Phi\left(y_{1}\right)\right) d y_{1}
\end{align*}
$$

where

$$
\phi(t) \equiv \frac{1}{\sqrt{2 \pi}} e^{-\frac{1}{2} t^{2}} \text { and } \Phi(t) \equiv \int_{-\infty}^{t} \phi(s) d s
$$

The first integral is

$$
[-\phi(t) \bar{\Phi}(t)]_{\theta}^{\infty}-\int_{\theta}^{\infty}(\phi(t))^{2} d t=\phi(\theta) \bar{\Phi}(\theta)-\int_{\theta}^{\infty}(\phi(t))^{2} d t
$$

the second integral is $\Phi(\theta)[-\phi(t)]_{\theta}^{\infty}=\Phi(\theta) \phi(\theta)$, and the third integral is

$$
[-\phi(t)(\Phi(\theta)-\Phi(t))]_{t=-\infty}^{\theta}-\int_{-\infty}^{\theta}(\phi(t))^{2} d t=-\int_{-\infty}^{\theta}(\phi(t))^{2} d t
$$

Hence the sum is

$$
\begin{equation*}
\phi(\theta)-\int_{\theta}^{\infty}(\phi(t))^{2} d t-\int_{-\infty}^{\theta}(\phi(t))^{2} d t \tag{5.2}
\end{equation*}
$$

and its derivative is

$$
-\theta \phi(\theta)+(\phi(\theta))^{2}-(\phi(\theta))^{2}=-\theta \phi(\theta)<0
$$

Therefore the sum is maximized at $\theta=0$, and has the value

$$
\phi(0)-\int_{0}^{\infty}(\phi(t))^{2} d t-\int_{-\infty}^{0}(\phi(t))^{2} d t=\frac{1}{\sqrt{2 \pi}}-2 \cdot \frac{1}{4 \sqrt{\pi}}=\frac{2-\sqrt{2}}{2 \sqrt{2 \pi}}
$$

(Note that $\left.\int_{0}^{\infty}(\phi(t))^{2} d t=\frac{1}{4 \sqrt{\pi}}\right)$. Thus we arrive at

Theorem 5 In Case $5 a$, the optimal choice is $\theta^{*}=0$, and the common OPR is $\frac{2-\sqrt{2}}{\sqrt{2 \pi}} \approx$ 0.2337 .

5b. Symmetric Exponential Distribution in $(-\infty, \infty)$. The p.d.f. is $f(x)=\frac{1}{2} e^{-|x|}$. The mean value is 0 . We consider the case $\theta \geq 0$ only. The reward I can get is shown by Figure 5, again.

I's total expected reward is $2 \times \frac{1}{4}$ times of

$$
\begin{align*}
& \int_{\theta<y_{1}<y_{2}<\infty} \int_{y_{1}} e^{-\left(y_{1}+y_{2}\right)} d y_{1} d y_{2}+\left\{\int_{0}^{\theta} e^{-y_{2}} d y_{2}+\int_{-\infty}^{0} e^{y_{2}} d y_{2}\right\}\left\{\int_{\theta}^{\infty} y_{1} e^{-y_{1}} d y_{1}\right\}  \tag{5.3}\\
& +\left\{\int_{-\infty<y_{1}<0<y_{2}<\theta} y_{1} e^{y_{1}-y_{2}} d y_{1} d y_{2}+\iint_{0<y_{1}<y_{2}<\theta} y_{1} e^{-\left(y_{1}+y_{2}\right)} d y_{1} d y_{2}\right. \\
& \\
& \left.+\int_{-\infty<y_{1}<y_{2}<0} \int_{1} e^{y_{1}+y_{2}} d y_{1} d y_{2}\right\}
\end{align*}
$$

i.e., the sum of the reward in $Q^{(1)}, Q^{(2)}$ and $Q^{(3)}$, in this order. For computation of these integrals, we used the formulas (4.3).

The part in $Q^{(1)}$ is

$$
\begin{aligned}
& \int_{\theta}^{\infty} e^{-y_{2}} d y_{2} \int_{\theta}^{y_{2}} y_{1} e^{-y_{1}} d y_{1}=\int_{\theta}^{\infty}\left\{(1+\theta) e^{-\theta}-\left(1+y_{2}\right) e^{-y_{2}}\right\} e^{-y_{2}} d y_{2} \\
&=\left(\frac{1}{2}+\theta\right) e^{-2 \theta}-\frac{1}{4}(1+\theta) e^{-\theta}
\end{aligned}
$$

the part in $Q^{(2)}$ is $(1+\theta)\left(2-e^{-\theta}\right) e^{-\theta}$, and the part in $Q^{(3)}$ is

$$
\begin{aligned}
-\left(1-e^{-\theta}\right) & +\int_{0}^{\infty}\left\{1-\left(1+y_{2}\right) e^{-y_{2}}\right\} e^{-y_{2}} d y_{2}+\int_{-\infty}^{0}-\left(1-y_{2}\right) e^{2 y_{2}} d y_{2} \\
& =e^{-\theta}-\int_{0}^{\infty}(1+t) e^{-2 t} d t-\left(\frac{1}{2}-\frac{1}{4}\right)=e^{-\theta}-1
\end{aligned}
$$

Hence the sum of the three parts is

$$
\begin{align*}
\left\{\left(\frac{1}{2}+\theta\right) e^{-2 \theta}-\frac{1}{4}(1+\theta) e^{-\theta}\right\}+(1+\theta)(2 & \left.-e^{-\theta}\right) e^{-\theta}+\left(e^{-\theta}-1\right)  \tag{5.4}\\
& =\left(\frac{11}{4}+\frac{7}{4} \theta\right) e^{-\theta}-\frac{1}{2} e^{-2 \theta}-1 .
\end{align*}
$$

Its derivative satisfies

$$
-e^{-\theta}\left(1+\frac{7}{4} \theta-e^{-\theta}\right)\left\{\begin{array}{l}
= \\
<
\end{array}\right\} 0, \text { if } \theta\left\{\begin{array}{l}
= \\
>
\end{array}\right\} 0
$$

Therefore we obtain
Theorem 6 In Case 5b, the optimal choice is $\theta^{*}=0$ and the common OPR is $2 \times \frac{1}{4} \times$ $\left(\frac{11}{4}-\frac{3}{2}\right)=\frac{5}{8}(=0.625)$.

## 6 Three Remarks

Remark 1 According to our theorems 1~6, player's optimal choice $\theta^{*}$ is equal to the expected value of each r.v.. The choice is made to the effect that the information concerning the covered r.v. obtained by the opponent, becomes least.

Suppose that player II doesn't employ any deception strategy i.e., he chooses $Y_{1}$ and $Y_{2}$ with probability $1 / 2$ each, and opens it. If player I knows this policy of his opponent, then player I can get $E_{f}\left(Y \vee \mu_{f}\right)$, where $\mu_{f}=E_{f}(Y)$. It is clear that

$$
E_{f}\left[Y \vee \mu_{f}\right]= \begin{cases}\int_{0}^{1}\left(y \vee \frac{1}{2}\right) d y=\frac{5}{8}, & \text { if } f(y)=1, \forall y \in[0,1] \\ \int_{0}^{\infty}(y \vee 1) e^{-y} d y=1+e^{-1} \approx 1.368, & \text { if } f(y)=e^{-y}, \forall y \in[0, \infty) \\ \int_{0}^{\infty} y \phi(y) d y=\frac{1}{\sqrt{2 \pi}} \approx 0.399, & \text { if } f(y)=\phi(y), \forall y \in(-\infty, \infty)\end{cases}
$$

these of which are grater than

$$
\frac{7}{12}\left(\text { in Th.1), } \frac{1}{2}+2 e^{-1} \approx 1.236\left(\text { in Th.3) and } \frac{2-\sqrt{2}}{\sqrt{2 \pi}} \approx 0.2337\right. \text { (in Th.5) }\right.
$$

respectively, in this order. We interprete that this means that II can deceive his opponent by enploying the deception strategy as mentioned Section 2. The situation is the same as for player I.

Remark 2 Let $\left(X_{1}, X_{2}\right)\left[\left(Y_{1}, Y_{2}\right)\right]$ be a bivariate correlated r.v. with p.d.f. $f\left(x_{1}, x_{2}\right)$ $\left[f\left(y_{1}, y_{2}\right)\right]$. For example

$$
\begin{equation*}
f\left(x_{1}, x_{2}\right)=1+\gamma\left(1-2 x_{1}\right)\left(1-2 x_{2}\right), \quad\left(x_{1}, x_{2}\right) \in[0,1]^{2} \tag{6.1}
\end{equation*}
$$

with $\gamma,|\gamma| \leq 1$, is a given constnt, and the correlation coefficient is $(1 / 3) \gamma$. This is one of the simplest one that has identical uniform marginals and correlated components. See Ref. $[3,5]$.
Also

$$
\begin{equation*}
f\left(x_{1}, x_{2}\right)=e^{-\left(x_{1}+x_{2}\right)}\left\{1+\gamma\left(2 e^{-x_{1}}-1\right)\left(2 e^{-x_{2}}-1\right)\right\}, \quad\left(x_{1}, x_{2}\right) \in[0, \infty)^{2} \tag{6.2}
\end{equation*}
$$

where $\gamma,|\gamma| \leq 1$, has identical exponential marginals $f(x)=e^{-x}$ and the correlation coefficient is $(1 / 4) \gamma$. This is one of the simplest one that has identical exponential marginals and correlated components.

Finally, bivariate normal distribution

$$
\begin{align*}
f\left(x_{1}, x_{2}\right)=\frac{1}{2 \pi \sqrt{1-\rho^{2}}} \exp \left[-\frac{1}{2\left(1-\rho^{2}\right)}\left(x_{1}^{2}-2 \rho x_{1} x_{2}+x_{2}^{2}\right)\right] &  \tag{6.3}\\
& \left(x_{1}, x_{2}\right) \in(-\infty, \infty)^{2}
\end{align*}
$$

where $\rho,|\rho|<1$, is the correlation coefficient. This has the identical marginal p.d.f. $\phi(x) \equiv$ $\frac{1}{\sqrt{2 \pi}} e^{-x^{2} / 2}$.

The two-sided games of deception for bivariate distributions (6.1), (6.2) and (6.3) are an interesting problem to be solved (see Ref.[5]).

Remark 3 We give an example of non-simple two-sided games of deception. Let $X_{1}$ and $X_{2}\left[Y_{1}\right.$ and $\left.Y_{2}\right]$ be i.i.d. r.v.s with p.d.f. $f(x)[g(y)] . f(x)$ and $g(y)$ are different p.d.f.s. Player I (II) chooses a number $\theta_{1}\left(\theta_{2}\right)$ and he opens the number nearest to $\theta_{1}\left(\theta_{2}\right)$ among $x_{1}, x_{2}\left(y_{1}, y_{2}\right)$, and covers the other number. Each player gets as his reward his opponent's $\left\{\begin{array}{c}\text { opened } \\ \text { covered }\end{array}\right\}$ number if it is $\left\{\begin{array}{l}>\theta_{1} \vee \theta_{2} \\ <\theta_{1} \wedge \theta_{2}\end{array}\right\}$ and $\frac{1}{2}$ (opened + covered), if otherwise. I and II want to choose $\theta_{1}$ and $\theta_{2}$, respectively, under which each player maxmizes the expected reward he can get. The three cases (1) $f(x) \equiv 1$ and $g(x)=2 x, x \in[0,1]$, (2) $f(x)=e^{-x}$ and $g(x)=x e^{-x}, x \in[0, \infty),(3) f(x)=\phi(x)$ and $g(x)=\frac{1}{2} e^{-|x|}, x \in(-\infty, \infty)$, would be interesting.

Remark 4 There are a number of interesting researches around the games of deception. Among these are Ref.[2, 3, 4, 6 and 7].

## References

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