Diversification Can Control Probability of Default or Risk, but Not Both

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Abstract: One of the controversies of diversification is that it may not be beneficial to banks, as it tends to increase systemic risk. Recent theoretical and empirical work have addressed this problem. We argue, from a theoretical perspective, that this controversy ultimately depends on how risk is assessed or measured. In particular, we observe that when one talks about random losses (risk) there are two intertwined approaches. On the one hand, one can fix the loss level and ask with what probability does that occur. On the other, one can fix a confidence level (or probability of loss) and ask, for example, what is the smallest loss with that probability. In a banking system, a systemic crisis occurs when all banks default simultaneously. Using the theoretical work of Wagner, where he proposed a simple model of a banking system in which a systemic crisis increases with diversification, we extend his analysis to show that if one allows for short positions; then the probability of default decreases, but the risk, measured by the value at risk (a non-coherent risk measure) increases. This brings up an interesting methodological question for risk management: Should we consider the probability of a given (acceptable) loss or, should we consider the minimum loss with an acceptable probability? We show that, within Wagner’s model and depending on which question is asked, a different answer can be obtained. This, in turn, lead us to discuss some implications of these results for risk managers and regulators.

Keywords: systemic crisis; risk measures; value at risk; short positions; perceptions of risk

1. Introduction

Recent theoretical (Allen and Carletti 2006; Allen and Gale 2005; Allen et al. 2012; De Young 2012; Ibragimov et al. 2011; van Oordt 2014; Wagner 2010) and empirical studies (Yang et al. 2020; Bégin et al. 2019; Slijkerman et al. 2013; De Jonghe 2010; Olibe et al. 2008) have addressed the problem of diversification and systemic risk. One common argument that can be found in the literature, both theoretical and empirical, is that as banks diversify their portfolios (although they might reduce the risk of individual failure) the systemic risk increases because banks become more similar to each other. In other words, by overlapping in related business activities, the probability of joint failure of the banking system as a whole increases. The recent financial crisis of 2007–2008 is usually used as an illustration of this point.

From a theoretical perspective, we believe that this argument has been presented in the most simple, and powerful way, in the work of Wagner (2010). The theoretical model that Wagner (2010) presents shows how diversification decreases the probability of failure for each bank individually, but it does not decrease the probability of default of the banking system as a whole (i.e., systemic risk). It must be said that an important, and very interesting, theoretical complement to this argument has been provided by Ibragimov et al. (2011). In this paper we focus solely in the work of Wagner (2010), because Ibragimov et al. (2011)
find this result in the case of full diversification whereas Wagner (2010) establishes this result for any degree of diversification.

The system in Wagner (2010) consists of two banks. Each bank collects a unit of funds from investors of which a share \( d \) is in the form of deposits. Bank 1 invests in asset \( X \) and bank 2 in asset \( Y \). The asset returns, \( x \) and \( y \), are identically and independent random variables, uniformly distributed in \([0, s]\). In the model, a bank defaults when banks cannot pay depositors, i.e., \( x < d \) or \( y < d \). The event \( \{x \leq d, y \leq d\} \) is called a systemic crisis, and Wagner noticed that if the banks diversify according to \( \nu(1) = (1 - r_1)x + r_1 y \) and \( \nu(2) = r_2 x + (1 - r_2) y \), it can be proved that:

\[
P(\nu_1 \leq d, \nu_2 \leq d) > P(x \leq d, y \leq d).
\]  

(1)

Since the probability for the non-diversified system is given by the (normalized) area of the square \( D \) in Figure 1 and the probability of joint default is given by that of the area of the combined regions \( H \cup D \cup M \), the validity of (1) is clear.

Even though the case depicted in Figure 1 is one of several possible cases considered in Wagner (2010), he considered diversified portfolios without short positions. One of the aims of this work is to extend Wagner’s theoretical analysis by including portfolios with short positions. We show that if we consider the generic case depicted in Figure 2, that when each bank diversifies by including short positions on the other bank’s portfolio; then the probability of joint default decreases, even though the probability of each bank defaulting individually may increase. The other aim of this work is to point out a dilemma that confronts the risk managers and/or regulators: Should one minimize the probability of collective loss or minimize the joint risk of default? We show that each choice leads to a different strategic decision.

![Figure 1. Long positions.](image1)

![Figure 2. Short positions.](image2)
The remainder of this work is organized as follows. In Section 2 we recall Wagner’s model. Since in that model the different probabilities of interest reduce to the computation of areas in a square of size $s$, we describe the areas determined by the different portfolios geometrically. In Sections 3 and 4 we consider the extensions of Wagner’s model. First, in Section 3 we prove that when diversification includes short positions and leverage, the probabilities of default decrease with diversification. In particular, the probability of joint default or systemic crisis decreases. In Section 4 we depart from Wagner’s model in that we consider portfolios with a return of arbitrary sign, and we shall see that in this model the probability of systemic crisis decreases when short positions are allowed. In Section 5 we regard the problem from the point of view of risk theory. We carry out an analysis using the value at risk (VaR) (which is a non-coherent risk measure). Since we are considering Gaussian returns, the VaR happens to be the standard deviation, and this makes the analysis simple. Clearly, if one were to consider coherent risk measures, the risk of a diversified portfolio would always be less than the sum of the risks of the individual banks. In Section 6 we collect some final remarks and discuss some implications of our results for risk managers and regulators. In Appendix A we offer details of computations presented in previous sections. This paper can be seen as an extension of (Cadenas et al. 2020); where short positions are not considered and there is no reference to Gaussian returns.

2. The Model

Using the same notations as in Wagner (2010), consider the financial positions of the two investors with diversified portfolios given by:

\[ v(1) = (1 - r_1)x + r_1y \]
\[ v(2) = r_2x + (1 - r_2)y \]

A bank run takes place whenever \( \{ v(1) < d \} \) or \( \{ v(2) < d \} \), which are the regions in the \( (X, Y) \) plane (respectively) bounded by the lines \( (1 - r_1)x + r_1y = d \) and \( r_2x + (1 - r_2)y = d \). Each bank is allowed to invest a share of their funds \( r_i \in [0, 1] \) \((i = 1, 2)\) in the other’s bank portfolio. In the case of short positions, we say that each bank is allowed to short its portfolio anywhere in the range given by \( r_i \in [-1, 0] \). After all, since the model assumes that each bank manages one unit of funds from risk-neutral investors, when considering short positions it is reasonable to assume that banks cannot short beyond one unit of funds.

Some Generic Observations about the Default Regions

In order to compute the probabilities of default, we begin by exploring the positioning of these lines as \( r \) varies. Denote by \( y_1(x) = (1 - 1/r_1)x + d/r_1 \) (resp. \( y_2(x) = -r_2/(1 - r_2)x + d/(1 - r_2) \)) the equation of these lines. When \( r_1 = 0 \) we have the vertical line (infinite slope) through \( x = d \) (resp. to \( r_2 = 0 \) we have the horizontal line \( y = d \)). The line \( (1 - r_1)x + r_1y = d \) rotates counterclockwise as \( r_1 \) increases from 0 to 1, and it rotates clockwise as \( r_1 \) decreases from 0 to \(-1\). To see this, note that the slope of this line is \( dy_1/dx = 1 - (1/r_1) \), thus the line is vertical for \( r_1 = 0 \) and increases from \(-\infty \) to 0 as \( r_1 \) increases from 0 to 1. Similarly, \( r_2x + (1 - r_2)y = d \) rotates clockwise as \( r_2 \) increases from 0 to 1, and it rotates counterclockwise as \( r_2 \) decreases from 0 to \(-1\). For \( r_1 < 0 \) the region \( \{ v_1 \leq d \} \) (resp. \( \{ v_2 \leq d \} \) for \( r_2 < 0 \)) is the region to the left (resp. to the right) of the line \( y_1(x) \) (resp. \( y_2(x) \)).

Finally, note that \( y_1 \) and \( y_2 \) rotate towards the diagonal as \( r_1, r_2 \to -\infty \). Let us suppose that \( r_1 < 0 \) and \( r_2 < 0 \) are such that the lines are as in Figure 2. The intercepts of the first line are \((d/(1 - r_1), 0)\) and \(((d - sr_1)/(1 - r_1), s)\) and those of the second line are \((0, d/(1 - r_2))\) and \((s, d - sr_2)\).
3. The Probabilities of Default in Wagner’s Model

3.1. The Probability of an Individual Bank Default

Here we want to compare \( P(v_1 \leq d) \) to \( P(x \leq d) \). That is, the effect of diversification (with shorting) to the first bank. The analysis for the second being similar and we shall skip it.

A simple computation of the area within \([0,s]^2\) to the right of the line \( y_1(x) \) is the area of the region below the graph of \( y_1(x) \) divided by \( s^2 \), that is:

\[
P(v_1 \leq d) = \frac{1}{s^2} \int_0^s \frac{d - sr_1}{1 - r_1} + s(s - \frac{d - sr_1}{1 - r_1}) = \frac{1}{2(1 + |r_1|)}(|r_1| + 2(1 - \frac{d}{s})). \tag{2}
\]

Here we put \( |r_1| = -r_1 \). To determine for what values of \( d/s \), is \( P(v_1 \leq d) \) smaller or larger than \( \frac{d}{s} = P(x \leq d) \); we rewrite this as

\[
\frac{1}{2(1 + |r_1|)}(|r_1| + 2(1 - \frac{d}{s})) \left( \begin{array}{c} \leq \frac{d}{s} \\ \geq \frac{d}{s} \end{array} \right) \Leftrightarrow \frac{d}{s} \geq \frac{r_1 + 1}{2} \frac{d}{s} (|r_1| + 2).
\]

That is, in case of default threshold \( d \) being bigger than half of the bank capital \( s/2 \), then any short position will decrease the probability of default. Otherwise, that is when \( d < s/2 \), diversification increases the probability of individual default. This is a rather curious role of the default threshold in this model.

We provide details of the derivation of Equation (2) in Appendix A.1.

3.2. The Probability of Systemic Crisis in Wagner’s Model with Shorting

We are now ready to compute the probability of a systemic crisis, that is the probability that both banks have wealth less than the critical value \( d \). We have already mentioned that Wagner (2010) established that the diversified portfolios with \( 0 \leq r_1 \leq 1 \) and \( 0 \leq r_2 \leq 1 \) are such that (1) holds.

When \( r_1 < 0 \) and \( r_2 < 0 \), which is when the situation in Figure 2 holds; we then have the following result.

**Theorem 1.** With the notations and under the assumptions introduced above, we get:

\[
P(v_1 \leq d, v_2 \leq d) = \frac{1 - (r_1 + r_2)/2}{(1 - r_1)(1 - r_2)} \leq P(x \leq d, y \leq d) = \left(\frac{d}{s}\right)^2. \tag{3}
\]

**Proof.** The computation leading to the first equality is sketched in Appendix A.2.

To verify the inequality we notice that if \( r_1, r_2 < 0 \) then \( (r_1 + r_2)/2 < 0 < r_1r_2 \), which in turn leads to

\[
\frac{1 - (r_1 + r_2)/2}{(1 - r_1)(1 - r_2)} \leq 1.
\]

The moral seems to be that if each bank can short their position in the other bank, in order to leverage their own assets, then the probability of systemic crisis decreases regardless of the size of the short positions assumed by both banks. This would require the existence of a third party to finance the leverage. \( \square \)

4. The Probabilities of Default in a Gaussian Model

Here we depart from Wagner’s model in the sense that \( x \) and \( y \) are considered to be the returns on the bank’s assets, and are supposed to be centered Gaussian random variables, with unit variance and correlation \( \rho \). That is, their joint distribution has a probability density

\[
f(x, y) = \frac{1}{2\pi(1-\rho^2)^{1/2}} \exp \left( -\frac{x^2 - 2\rho xy + y^2}{(1-\rho^2)} \right).
\]
In this case, the range of \((x, y)\) is the whole plane. Once an admissible default threshold \(d < 0\) is agreed upon, the joint default region is the "orthant" \(\{(x, y) : x \leq d, y \leq d\}\). If we consider the diversified portfolios \(v(1) = (1 - r_1)x + r_1y\) and \(v(2) = r_2x + (1 - r_2)y\), then the geometry of the default regions is as above, the lines that define the boundary of the default regions extend to infinity in both directions. At this point let us only note that the probability of loss of any of the banks is \(P(x \leq d) = \Phi(d)\), where \(\Phi\) stands for the cumulative distribution function of the standard normal random variable.

It is also clear that when \(r_1 > 0\) and \(r_2 > 0\), Wagner’s observations continue to be valid, that is, Equation (1) holds true in this model as well. Let us now examine the effect of diversification depending on the type of diversification or leveraging.

### 4.1. The Probabilities of Loss of a Diversified Bank

The Gaussian model allows us to compute the probability of loss of the diversified portfolio. Note that \(v_1 = (1 - r_1)x + r_1y\) is a centered Gaussian random variable with variance \(\sigma^2(r_1) = (1 - r_1)^2 + 2\rho r_1 (1 - r_1) + r_2^2\). Therefore

\[
P(v_1 \leq d) = P\left(\frac{v_1}{\sigma(r_1)} \leq \frac{d}{\sigma(r_1)}\right) = \Phi\left(\frac{d}{\sigma(r_1)}\right).
\]  

Since \(\sigma(r_1) \leq 1\) or \(\sigma(r_1) > 1\) depending on whether \(0 \leq r_1 \leq 1\) or not, we have

**Theorem 2.** With the notations introduced above, it follows from (4) that:

\[
P(v_1 \leq d) \begin{cases} 
\geq P(x \leq d) & \text{for } 0 \leq r_1 \leq 1 \\
< P(x \leq d) & \text{otherwise}.
\end{cases}
\]  

That is, shorting or leveraging decreases the probability of default for an individual bank.

The role that the standard deviation plays in this analysis, will reappear when we examine two standard risk measures, namely the VaR and the expected shortfall.

### 4.2. The Probability of Systemic Crisis

In the extended case, the region \(S\) of default, that is \(\{v_1 \leq d, v_2 \leq d\}\) is the region between the lines \(y_1(x)\) and \(y_2(x)\) contained in the "orthant" determined by the half planes \(\{x \leq d\}\) and \(\{y \leq d\}\) as depicted in Figure 3.

![Figure 3. Region of systemic crisis for general returns.](image)
As the point \((d, d)\) is on both lines as well as on the lines \(\{x = d\}\) and \(\{y = d\}\), as above, we might think of \(y_1\) as obtained by rotating \(\{x = d\}\) clockwise about \((d, d)\), and \(y_2\) is obtained rotating \(\{y = d\}\) counterclockwise about \((d, d)\). To be more precise \(S\) in the 4th orthant contained between the lines defined by the angles:

\[
\theta_1 = \arctan\left(1 - \frac{1}{r_1}\right) > \theta_2 = \arctan\left(-\frac{r_2}{1-r_2}\right)
\]

when the origin of the coordinate system is put at \((d, d)\). Note that when \(r_1 \uparrow 0\) then \(\theta(r_1) \to 3\pi/2\), and as \(r_2 \uparrow 0\) then \(\theta(r_2)\) decreases to \(\pi\). That is \(S\) tends to the region \(\{x \leq d, y \leq d\}\). We now clearly have:

**Theorem 3.** With the notations just introduced, when \(r_1 < 0\) and \(r_2 < 0\), the probability of joint default of the diversified portfolios satisfies:

\[
P(v_1 \leq d, v_2 \leq d) < P(x \leq d, y \leq d).
\]

As \(r_1 \uparrow 0\) and \(r_2 \uparrow 0\)

\[
P(v_1 \leq d, v_2 \leq d) \uparrow P(x \leq d, y \leq d).
\]

**Proof.** In principle \(P(v_1 \leq d, v_2 \leq d)\) can be computed as

\[
P(v_1 \leq d, v_2 \leq d) = P((x, y) \in S) = \int_S f(x, y)\,dx\,dy.
\]

The problem is that the probability density of \((x, y)\) is not invariant under translations nor under rotations, therefore, to go beyond this generic statement, we now suppose that \(d = 0\) (that is, in this model we count any negative return as a loss). If we introduce polar coordinates \(x = r \cos \gamma\) and \(y = r \sin \gamma\), the probability \(P((x, y) \in S)\) can be computed as

\[
P((x_1, y_1) \in S) = \frac{1}{2\pi} \frac{1}{1 - \rho^2} \int_0^{\theta_1} \int_0^{\theta_2} e^{-\frac{1}{2(1 - \rho^2)} r^2(1 - \rho \cos 2\gamma)} \,r\,dr\,d\gamma.
\]

Integrating with respect to \(r\) we obtain

\[
P((x_1, y_1) \in S) = \frac{1}{2\pi} \int_0^{\theta_1} \int_0^{\theta_2} e^{-\frac{1}{2(1 - \rho^2)} r^2(1 - \rho \cos 2\gamma)} \,d\gamma = \frac{1}{2\pi} \int_0^{\theta_2} \frac{d\gamma}{1 - \rho \cos \gamma}.
\]  \hspace{1cm} (6)

We collect these remarks under the following statement.

**Theorem 4.** With the notations introduced above, the probability of systemic crisis is given by (6).

**Comments**

(a) Note that \(P((x_1, y_1) \in S)\) is a increasing function of the coefficient of correlation \(\rho\).

(b) When \(\rho = 0\), (the case considered by Wagner)

\[
P((x_1, y_1) \in S) = \frac{1}{2\pi} (\theta(r_1) - \theta(r_2)).
\]

(c) As \(r_1 \to 0\) and \(r_2 \to 0\), then since \(\theta(r_1) \to 3\pi/2\) and \(\theta(r_2) \to \pi\), in this particular case:

\[
P((x_1, y_1) \in S) \to \frac{1}{4} = P(x \leq 0)P(y \leq 0)
\]
5. Risk Analysis

In this section, we shall verify that as \( r_1 < 0 \) then the risk of \( \nu_1 \) becomes larger than the risk of \( X \) when both are measured by the value at risk. Similarly, if we consider the total, or aggregated risk of the two diversified banks, the risk of the combined banks is larger than that of the combined non-diversified banks.

The value at risk (VaR) of a random asset of return \( W \), at confidence level \( \alpha \), modeled by a continuous random variable, in units of the initial investment, is given by the solution (in \( V \)) of

\[
P(W \leq -V) = 1 - \alpha.
\]

It is known that VaR is non-necessarily a coherent risk measure, despite the fact that it is enforced by regulators, and that it is very easy to estimate and interpret.

It is a standard exercise to verify that when \( W \) is a centered Gaussian random variable:

\[
VaR_\alpha(W) = \Phi^{-1}(\alpha)\sigma(W) \tag{7}
\]

5.1. The Risk of a Diversified Portfolio

The diversified portfolio \( \nu_1 = (1 - r_1)x + r_1 y \) is a centered Gaussian with variance \( \sigma^2(r_1) = (1 - r_1)^2 + 2r_1(1 - r_1) + r_1^2 \). Thus, according to (7), the value at risk of \( \nu_1 \) is

\[
VaR_\alpha(\nu_1) = \Phi^{-1}(\alpha)\left((1 - r_1)^2 + 2r_1(1 - r_1) + r_1^2\right)^{1/2} = \Phi^{-1}(\alpha) = VaR_\alpha(x)
\]

whenever \( r_1 < 0 \) (and of course for \( r_1 > 1 \)). That is, if we measure risk with VaR, entering a short or a leveraged position, increases the risk even though it decreases the probability of loss.

5.2. The Aggregated Risk

If we model the financial positions of the banking system by \( R = \nu_1 + \nu_2 \), in the same way that an individual bank would aggregate its risks, and we decide to measure the systemic risk by the value at risk, then, invoking (7) once more we obtain:

\[
VaR_\alpha(\nu_1 + \nu_2) = VaR_\alpha(\nu_1) = \Phi^{-1}(\alpha)\left((1 - r_1 + r_2)^2 + 2\rho(1 - r_1 + r_2)(1 - r_2 + r_1) + (1 - r_2 + r_1)^2\right)^{1/2} = \Phi^{-1}(\alpha)\sigma_{\text{div}}.
\]

This should be compared with the aggregated risk without diversification:

\[
VaR_\alpha(x + y) = \Phi^{-1}(\alpha)\sigma(x + y) = \Phi^{-1}(\alpha)\left(2(1 + \rho)\right)^{1/2} = \Phi^{-1}(\alpha)\sigma_{\text{tot}}
\]

and the conclusion is clear. We sum up these computations as follows.

**Theorem 5.** If we we aggregate the risk in the system as

\[
\nu_1 + \nu_2 = (1 - r_1 + r_2)x + (1 - r_2 + r_1)y
\]

then the value at risk at confidence level \( \alpha \) satisfies

\[
VaR_\alpha(\nu_1 + \nu_2) > VaR_\alpha(x + y).
\]
Notice that as far as the total loss being more than $2d$, we have

$$P(v_1 + v_2 \leq 2d) \geq P(x + y \leq 2d).$$

**Proof.** The proof of the first part was established prior to the statement. It is a reflection of the fact that VaR may not decrease with diversification. As far as the proof of the second assertion, notice that on the one hand $\sigma_{agg} \leq \sigma_{tot}$, and on the other

$$P(v_1 + v_2 \leq 2d) = \Phi\left(\frac{2d}{\sigma_{div}}\right) \quad \text{and} \quad P(x + y \leq 2d) = \Phi\left(\frac{2d}{\sigma_{tot}}\right).$$

Note that $\sigma_{tot} = \sigma_{div}$ whenever $r_1 = r_2$, that is when both banks are equally diversified. In this case, the probability of loss in the aggregated portfolio does not depend on the diversification. $\square$

6. Discussion

One of the controversies of diversification is that, although it may be beneficial for a bank in reducing the risk of its portfolio, it comes at the expense of increasing the likelihood of systemic risk. As banks diversify their portfolios, their business activities overlap and their portfolios become more similar to each other. This situation tends to expose the entire system to the same risks, and as a consequence, the probability of joint failure increases. In this line of reasoning, we extend the analysis by showing that when the returns of the bank’s assets are considered to be centered Gaussian random variables, as opposed to behaving under a uniform distribution as assumed by Wagner (2010), it continues to hold that diversification tends to increase systemic risk. However, a more important contribution of the paper is to show that when short positions are allowed; then the same measure of risk used by Wagner (2010) ends-up offering a different picture of diversification. That is to say, when short positions are allowed, diversification does not increase the probability of joint failure (i.e., systemic risk).

Furthermore, we show that not only the picture of diversification and systemic risk changes when short positions are allowed, but that the picture also changes when a different risk measure, like VaR, is used. We establish that when using VaR for evaluating short positions the risk of individual failure increases (rather than decreasing) with diversification; and also that the systemic risk decreases (rather than increasing) with diversification. These are the opposite conclusions than what we get when the risk measure is defined as in Wagner (2010). We would like to point out that one possible direct application of the idea of short positions could be the case of Government financial bailouts.

Our findings suggest a methodological problem, which consists of figuring out which of the two risk measures should be used and why. On one hand, we can pre-assign a tolerable loss threshold and then determine a portfolio that minimizes the probability of falling below the loss level. Or, on the other, we can fix a frequency of losses and then choose a portfolio that minimizes the loss with that frequency. Although these two procedures seem both to be equally reasonable; they can imply different views on diversification and, correspondingly, they assess individual and systemic risk in a different manner.

We don’t offer an ultimate solution to this problem here, but the finding in itself is relevant. Why? Because ambiguities of this sort generate a situation in which risk managers and regulators perceive risk in a manner that leads to opposite conclusions. Our findings are also relevant because we show that risk assessment changes depending on whether banks can assume a short position (capital leverage) or not. In addition, our results are suggestive of a host of problems that can be found in the literature about differences in risk perception in areas such as public health and tourism (Wolff et al. 2019; Renn 2004; Cui et al. 2016; Neuburger and Egger 2020), but for the case of banks. In the case of public health (Renn 2004), the discussion usually revolves around communication, conflict resolutions, and delineating the different rationale behind risk measures; which in the case of the banking system will not be much different.
Finally, and perhaps most importantly, the results we have presented here are also suggestive of problems about differences in risk perception that are discussed in the behavioral finance literature (Vlaev et al. 2009; Ricciardi 2004; Aren and Zengin 2016). Our work can be extended in various directions. One line of inquiry is to investigate whether allowing the system to consist of more than two banks would in any way change the conclusions. We are currently pursuing this line of work, following (van Oordt 2014), for the case in which securitization is allowed. Another line of inquiry might be to study how results would change in the case of flat tails distributions. A third line of inquiry can be to work, with one of the risk measures as a reference, the implications to diversification results would change in the case of flat tails distributions. A third line of inquiry can be to work, with one of the risk measures as a reference, the implications to diversification

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Appendix A

Appendix A.1. Derivation of (2)

The area of the triangle below curve \( y_1(x) = d/r_1 + (1 - 1/r_1)x \). Recall that \( r_1 < 0 \).

The area has two pieces: one triangle with basis \((d - s r_1)/(1 - r_1) - d/(1 - r_1) = |r_1|s/(1 + |r_1|)\) (since \(-r_1 = |r_1|\), and height \(s\). Then, a parallelogram with basis \( s - (d - s r_1)/(1 - r_1) = (s - d)/(1 + |r_1|)\) and height \(s\). Thus the area of the figure below the line \(y_1(x)\) is

\[
\frac{s^2}{2} \left( \frac{|r_1|}{1 + |r_1|} + \frac{1 - d/s}{1 + |r_1|} \right).
\]

Therefore

\[
P(v_1 \leq d) = \frac{1}{2} \left( \frac{|r_1| + 2(1 - d/s)}{1 + |r_1|} \right).
\]

Since \(|r_1|\) can be as close to 0 as we want, we need \(1 > d/s\).

Appendix A.2. Derivation of (3)

The area of the region \( S \) in Figure 2 is \(d^2\) minus the area of the two triangles bordering it within the square, that is, minus the sum:

\[
\frac{1}{2} d \left( \frac{d}{1 - r_1} + \frac{d}{1 - r_2} \right).
\]

That is

\[
\frac{d^2}{2} \left( 1 - \frac{1}{2(1 - (1 - r_1))} - \frac{1}{2(1 - (1 - r_1))} \right) = \frac{1 - (r_1 + r_2)/2}{(1 - r_1)(1 - r_2)}.
\]
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