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Kinship and Financial Networks, Formal Financial Access, and Risk Reduction[†]

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Many risks are present in rural developing economies: illness, weather, the sudden need to finance an investment opportunity, etc. Yet for many households in rural developing economies, consumption and investment are insured against short-term, idiosyncratic risks to a large extent, despite limited availability of formal banking and insurance products. The importance of both kinship networks and financial institutions in facilitating consumption smoothing and investment financing has been demonstrated in many settings. Yet, while the importance of kinship networks and financial access are each increasingly well-documented, the *channels* through which these effects occur and the relationship between kinship networks and financial access are not well understood. We use unique data from rural Thai households to examine this interplay.

I. Data

A. Household Data

Data are from the 1999–2005 monthly waves of the Townsend Thai Monthly Survey (Townsend et al. 1997). A total of 531 households in 16 villages are observed in each of the 84 months; we focus on this sample. Data were collected on households' demographic

composition, expenditure, and income. The most common occupation in the sample is rice farming (35 percent of households), followed by nonagricultural labor (including owning a nonagricultural business) (12 percent), growing corn (10 percent), raising livestock (9 percent), and agricultural wage labor (5 percent). Growing other crops, raising fish or shrimp, growing orchard crops, and construction each account for less than five percent.¹

B. Financial Network Data

Transfers with other households in the village are prevalent: gifts from other households in the same village equal nine percent of average expenditure. Borrowing from and lending to other households in the village are also widespread in the data. We use data on loans and transfers with other households in the village to construct a financial network of the village. For borrowing/lending and transfers with other households in the village, the surveyed household is asked to identify the structure (essentially, the address) in which the counterparty household lives. This is matched to a village census that records the address of every household in the village, and which is updated when households move. This allows us to identify the counterparty household for each within-village transaction, even if they are not themselves in the survey.² Some households are directly connected to banks, while others are indirectly connected, because they

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¹ See Samphantharak and Townsend (2010) for summary statistics and details of variable construction.

² We can observe links of the form $A \rightarrow B \rightarrow C$ even if B is not in the survey. We will miss links of the form $A \rightarrow B \rightarrow C \rightarrow D$ if neither B nor C is in the survey. This can cause nonclassical measurement error, causing some linked individuals to appear unlinked (Chandrasekhar and Lewis 2011), biasing the estimated effect of indirect links toward zero.

borrow from an individual who in turn borrows from a bank.

Because we are interested in the role of indirect access to financial institutions in facilitating access to credit, we construct directed links, from lender/giver to borrower/receiver. We have time-varying information on when households borrow from each other, but anticipation of being able to borrow may matter for consumption and investment decisions even in months when borrowing does not take place. Moreover, capital does not necessarily flow instantaneously through the financial network: a household may borrow from a bank in January, and then on-lend some of the money in March, for example. Therefore, we collapse the time variation in the data and construct, for each pair of households i, j in the dataset, an indicator for whether i ever borrows or receives transfers from j . We can then construct a variable γ_{ij} that represents the length of the shortest directed path from i to j . In network theory, this is referred to as the geodesic distance from i to j . Household i is said to be *reachable* by household j ($r_{ij} = 1$) if there exists any path from i to j ($\gamma_{ij} < \infty$).

For the 411 households who ever borrow or receive gifts from another household in the village, the average household borrows/receives from 3.2 other households (minimum 1, maximum 17). The average total amount borrowed from other households in the village over the 7-year sample, conditional on ever borrowing, is 73,727 baht.³ The average amount borrowed per transaction is 12,200 baht, which is equal to 60 percent of average monthly household expenditure. Thus, intravillage borrowing transactions tend to be large, but relatively infrequent, with the average household who ever borrows borrowing from other villagers 4.75 times over 84 months.

We also have information on borrowing from financial institutions. The institutions we consider here are commercial banks and the Bank for Agriculture and Agricultural Cooperatives (BAAC), which we refer to jointly as banks. We define $\gamma_{i,B}$ as the length of the shortest directed path from i to a bank: 1 if i borrows directly from the bank, 2 if i borrows from someone who

borrows from the bank, etc. Let $d_{i,B} = 1$ if the household is directly connected ($\gamma_{i,B} = 1$) and $r_{i,B} = 1$ if there exists any path to i from the bank ($\gamma_{i,B} < \infty$).

C. Kinship Network Data

We have data on the location of the parents, siblings, adult children, and parents' siblings of each surveyed household head and his/her spouse, if these relatives are living. If any of these relatives live in the same village as the surveyed household, we define the household as having kin in their village, $k_i = 1$. Otherwise, $k_i = 0$. Seventy-four percent of households have at least one relative living in the same village.

II. Empirical Specifications

A. Consumption

To investigate the impact of both kin and financial networks on consumption smoothing, we run regressions that modify the standard omnibus insurance specification (Townsend 1994) to allow the effect of income fluctuations to depend on the presence of kin, on net worth, and on direct and indirect connections to financial institutions. Alem and Townsend (2011) show that, with endogenous financial participation, a per-period shock common to all households who participate in the financial system should be added to the standard full insurance regression. Our notion of access to the financial system is connection (direct or indirect) to either the BAAC or to commercial banks. Therefore, our consumption-smoothing specification takes the form

$$(1) \quad \Delta c_{ivt} = \alpha_1 \Delta y_{ivt} + \alpha_2 \Delta y_{ivt} \times d_{i,B} \\ + \alpha_3 \Delta y_{ivt} \times r_{i,B} + \alpha_4 \Delta y_{ivt} \times k_i \\ + \alpha_5 \Delta y_{ivt} \times \bar{w}_i + \delta_{B,t} + \varepsilon_{it},$$

where c_{ivt} and y_{ivt} are, respectively, the per capita consumption and income of household i in month t , $d_{i,B}$ and $r_{i,B}$ indicate, respectively, direct and any connection to the financial system; k_i is an indicator for presence of kin in the village, \bar{w}_i is household i 's average net worth over the sample period, and $\delta_{B,t}$ is a common

³ This and all following references to baht refer to 2002 baht. The exchange rate in 2002 was approximately 42 baht to \$1.

time effect for all households connected to the financial system.

First-differencing removes any nontime varying characteristics of households that might be correlated with their ability to smooth consumption. For this reason, we do not include the main effects of financial access, presence of kin, or net worth.

B. Investment

To investigate the impact of kinship networks and financial networks on the ability to smooth investment in the face of cash flow fluctuations, we run regressions that modify the standard cash-flow sensitivity specification to allow the effect of income fluctuations to depend on the presence of kin, on net worth, and on connections to financial institutions. Alem and Townsend (2011) show that the investment and income variables should be scaled by total household assets to create an appropriate linear approximation to the optimal investment function of a firm. Because this will introduce heteroskedasticity, we compute heteroskedasticity-robust standard errors. We focus on positive investment events, and examine how the size of such events responds to the household's cash flow. We do not include household fixed effects in the investment regression because the number of positive investment events is small for each household. We include village-fixed effects, δ_v , to capture common characteristics such as suitability of the area for different occupations (rainfall, proximity to large towns, etc.), as well as a common time effect for all households connected to the financial system, $\delta_{B,t}$. We focus here on the effect of being connected at any distance, $r_{i,B}$. Thus our investment-smoothing specifications takes the form

$$\begin{aligned}
 (2) \quad \left(\frac{I}{A}\right)_{ivt} &= \alpha_1 \left(\frac{y}{A}\right)_{ivt} + \alpha_2 \left(\frac{y}{A}\right)_{ivt} \\
 &\quad \times r_{i,B} + \alpha_3 \left(\frac{y}{A}\right)_{ivt} \times k_i \\
 &\quad + \alpha_4 \left(\frac{y}{A}\right)_{ivt} \times \bar{w}_i + \beta_1 r_{i,B} \\
 &\quad + \beta_2 k_{i,B} + \beta_3 \bar{w}_i + \delta_v \\
 &\quad + \delta_{B,t} + \varepsilon_{it}.
 \end{aligned}$$

III. Results

A. Consumption

Due to space constraints, we do not present the results of estimating equation (1) in table form, but we discuss them here. First, we estimate a restricted version of equation (1) that does not allow the effect of income fluctuations to vary by financial access, kinship, or net worth. The results show that the Thai households in our sample achieve quite good consumption smoothing on average, with a one baht income change associated with a 0.0078 baht consumption change; however, this is significantly different from 0 at the 1 percent level, indicating that the households are not fully insured. Estimating a full version of equation (1), we see that households not connected at all to a bank are much worse insured than the average, with a 1 baht income change associated with a 0.1645 baht consumption change (significant at 1 percent) for this group. Being directly connected to a bank reduces the consumption-income comovement by 0.1658 baht (significant at 1 percent), yielding a net sensitivity of -0.0013 , insignificantly different from 0 ($p = 0.696$). An indirect connection has a virtually identical impact, reducing the consumption-income comovement, relative to no connection, by 0.1643 baht (significant at 1 percent), yielding a net sensitivity of 0.0002, insignificantly different from 0 ($p = 0.958$). Net worth is associated with significantly reduced consumption-income sensitivity, as expected, but the impact is small: one million baht in additional net worth is associated with a reduction in the consumption response to a 1 baht income change of 0.00021 baht (significant at 1 percent). Conditional on financial access and net worth, the effect of kin is to *increase* consumption sensitivity by 0.0102 baht per 1 baht income change (significant at 1 percent).

These results indicate that access to the formal financial system plays an important role in smoothing consumption in the face of income shocks. Strikingly, an indirect connection is as effective as a direct connection, suggesting that borrowing and lending among households acts to distribute capital from formal financial institutions. Ignoring the effect of being indirectly connected to financial networks and institutions, and using households not directly connected as a

TABLE 1—KINSHIP, FINANCIAL ACCESS, AND INVESTMENT

	No controls (1)	All house- holds (2)	Above- median investment size (3)	Below- median investment size (4)
Income	0.1078* (0.0649)	0.6526*** (0.1950)	0.6370*** (0.2102)	0.0077 (0.3359)
Income				
X Any link to bank		-0.1268 (0.1288)	-0.0821 (0.1292)	0.2931 (0.3983)
X Kin in village		-0.4136*** (0.1549)	-0.5056*** (0.1599)	0.4543 (0.3256)
X Net worth (mill. baht)		-0.1087 (0.0762)	-0.0405** (0.0205)	-0.3710 (0.2357)
Observations	6,055	5,794	2,319	3,463

Note: Heteroskedasticity-robust standard errors in parentheses.

***Significant at the 1 percent level.

**Significant at the 5 percent level.

*Significant at the 10 percent level.

comparison group, may yield biased estimates of the effect of financial access, due to the spillover of indirect access through other households.⁴

B. Investment

We now turn to discussing the results for smoothing investment in the face of cash flow fluctuations. Table 1 presents the results. Column 1 shows results for the full sample: unconditionally, a 1 baht increase in cash flow increases investment by 0.1078 baht, consistent with the findings of Samphantharak and Townsend (2010, chapter 6). Column 2 adds controls for kinship, financial access, and net worth (main effects and interactions with income; we report only the interactions with income to save space). Investment is highly sensitive to cash flow for households without kin in the village, with a 1 baht income change associated with a 0.6526 baht investment change, significantly different from 0 at the 1 percent level. The presence of kin in the village substantially mitigates this sensitivity, however, reducing the response to a 1 baht change by 0.4136 baht. Bank connections do not appear to be significantly helpful in smoothing investment, in contrast to their central role in consumption smoothing.

⁴ This echoes the findings of Angelucci and De Giorgi (2009) on the spillover effects of cash transfers in the presence of village-level insurance.

Why are consumption and investment different? The theory of the role of social networks suggests an explanation. Ambrus, Mobius, and Szeidl (2010) and Karlan et al. (2009) argue that, in the absence of formal commitment, networks that generate the most surplus for their members can sustain the largest flows of funds. For a household that has borrowed and now must repay, or that received insurance-motivated transfers and now must reciprocate, the threat of losing a high-value relationship, or seeing a friend or relative ostracized in response to the household's defection, relaxes the temptation to renege on their obligation. Anticipating this, households with strong ties can credibly transfer larger sums among each other.

Therefore, if the role of kin is to facilitate borrowing large amounts for investment, loans that borrowers could not otherwise commit to repay, we should see the effect of kin concentrated among households for whom investment opportunities are large relative to wealth. Since observed investment sizes are endogenous with respect to the household's access to financing, we use a household's occupation, in essence, as a proxy for the average scale of investment opportunity a given household might face. Our theory predicts that households in occupations where the average investment size is large relative to average wealth should derive the most benefit from the presence of kin. We group together the occupations with above-median observed investment-to-net worth ratios: business owners; farmers of crops other than rice, corn, and orchard trees; and nonagricultural workers (including business owners). The occupations with below-median investment-to-net worth ratios are rice farmers; farmers raising pigs and cows; corn and orchard tree farmers; and shrimp and fish farmers. Columns 3 and 4 of Table 1 present the results. As in column 2, the effect of cash flow fluctuations is allowed to vary by kinship, net worth, and connection to banks. Strikingly, it is for the occupation group with above-median ratios of observed investment to net worth that the effect of kin presence is evident: in this group, those without local kin experience an investment change of 0.637 associated with a 1 baht income change, and having kin in the village reduces this by 0.506 baht (significant at 1 percent). For occupation categories with smaller investment-to-net worth ratios, the effect of kin presence is small in magnitude and insignificant.

IV. Discussion and Conclusions

These results shed light on the question of *why* kinship networks and financial access matter in smoothing consumption and investment in the face of income volatility, by examining which type of networks (kin versus financial) matter for which type of insurance: the relatively small deviations of realized income from desired contemporaneous consumption, versus the potentially large difference between the scale of an investment opportunity and the amount of cash on hand to finance it. The fact that access to financial institutions appears to be helpful in smoothing consumption, while kinship networks are not helpful, suggests that financing needs of these magnitudes can be met most effectively with borrowing that can be collateralized implicitly or explicitly with tangible assets or threatened loss of participation in the financial network. On the other hand, kinship networks are important in financing investment for transactions too large to be collateralized with tangible assets, so that extended or nonpecuniary punishments by kin are important in assuring lenders that their loans will be repaid.

Our finding that being indirectly connected to the financial system is as beneficial as a direct connection implies that not every household in a village needs to use the banking system directly in order to benefit, if interpersonal gifts and lending are widespread. It also suggests that evaluating financial access by comparing those who use the banking system to those who do not may yield significant misestimates of the effect of financial access. Those without direct or indirect access and those without kin in a village remain highly vulnerable to fluctuations, while the connected and those with kin do relatively well.

REFERENCES

- Alem, Mauro, and Robert M. Townsend.** 2011. "An Evaluation of Financial Institutions: Impact on Consumption and Investment Using Panel Data and the Theory of Risk-Bearing." Unpublished.
- Ambrus, Attila, Markus Mobius, and Adam Szeidl.** 2010. "Consumption Risk-Sharing in Social Networks." Unpublished.
- Angelucci, Manuela, and Giacomo De Giorgi.** 2009. "Indirect Effects of an Aid Program: How Do Cash Transfers Affect Ineligibles' Consumption?" *American Economic Review* 99 (1): 486–508.
- Chandrasekhar, Arun G., and Randall Lewis.** 2011. "Econometrics of Sampled Networks." Unpublished.
- Karlan, Dean, Markus Mobius, Tanya Rosenblat, and Adam Szeidl.** 2009. "Trust and Social Collateral." *Quarterly Journal of Economics* 124 (3): 1307–61.
- Samphantharak, Krislert, and Robert M. Townsend.** 2010. *Household as Corporate Firms: An Analysis of Household Finance Using Integrated Household Surveys and Corporate Financial Accounting*. New York: Cambridge University Press.
- Townsend, Robert M.** 1994. "Risk and Insurance in Village India." *Econometrica* 62 (3): 539–91.
- Townsend, Robert M., Anna L. Paulson, Sombat Sakuntasathien, Tae Jeong Lee, and Michael Binford.** 1997. "Questionnaire Design and Data Collection for NICHD Grant 'Risk, Insurance and the Family' and NSF Grants." Unpublished.