Teams, Location, and Productivity^{*}

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ABSTRACT

The economics literature highlights a substantial influence of larger cities in improving worker skills and productivity, in particular, due to better means of information dissemination and learning. Studies in a variety of disciplines also reveal a positive role of team management for firm performance that results from diversified knowledge and skill sets of the team members. This suggests a positive relation between performance gains associated with team management and city size. Indeed, using U.S. equity mutual fund and demographic data, we show that teambased managerial approach leads to drastically different gains across different locations. We find that on average team-managed funds significantly outperform single-managed ones only in larger metropolitan areas. This result holds for an array of fund performance metrics and is robust to the inclusion of fund and manager characteristics controls. The effect, however, is concentrated among portfolio managers that survive several years of collaboration with the same team members. We interpret our findings as evidence of the enhancement of information and knowledge spillover in larger cities within team-based organizational structures.

JEL classifications: D70, G23, J24

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1. Introduction

This paper links urban economics to organizational structure. Specifically, it examines the performance of team-managed and single-managed mutual funds across U.S. metropolitan areas. The motivation for this project is two-fold and is based on the current developments in economics and organizations.

The urban agglomeration literature generally finds strong empirical evidence of higher worker productivity in larger cities relative to small ones (e.g., see Wheeler, 2001; Rosenthal and Strange, 2004; Chrisoffersen and Sarkissian, 2009; Combes, et al., 2012).¹ One theory that supports this finding is proposed by Helsley and Strange (1990, 1991) who consider larger cities as being more attractive for skilled workers, because their skills can be better matched with jobs in more populous areas. Another theory, proposed by Jacobs (1969) and further developed by Lucas (1988), Glaeser, Kallal, Scheinkman, and Shleifer (1992), and Glaeser (1999), considers large cities to be more conductive for learning and knowledge transfers that help enhance firm productivity.²

Academic sources also discuss possible benefits of group decision-making for firm output. For instance, Sharpe (1981), Barry and Starks (1984), and Sah and Stiglitz (1986, 1991) argue that portfolio management teams achieve diversification of style and judgment, thus reducing portfolio risk and improving performance. Many empirical studies support the opinion and risk diversification theories in groups.³ However, only a couple of papers find direct evidence that teams improve on firm productivity. Hamilton, Nickerson, and Owan (2003) find that teams increase productivity, and that this increase is more apparent among high-ability workers. Patel and Sarkissian (2016) show that team-managed mutual funds outperform their

¹ The earliest reference to this phenomenon can be found in Adam Smith (1776).

 $^{^{2}}$ An alternative explanation for higher productivity in large cities, due to Melitz (2003), uses the premise that larger number of firms in the area increases competition and, therefore, drives less productive firms out. However, mutual funds compete with each other far beyond their respective locations. In addition, Combes, et al. (2012) find no support for this explanation.

³ Studies in this area are based on signaling games experiments, such as Cooper and Kagel (2004), Blinder and Morgan (2005), Feri, Irlenbusch, and Sutter (2010), as well as on work of Barber, Heath, and Odean (2003), Adams and Ferreira (2010), Bar, Kempf, and Ruenzi (2011), and many others.

single-managed peers across a variety of performance evaluation measures. Yet, other studies point out to certain negative consequences of collective decision-making, such as the problems of moral hazard and free-riding (e.g., Alchian and Demsetz, 1972; Holmstrom, 1982; Nalbantian and Schotter, 1997).

Thus, since urban agglomeration theories suggest the existence of more knowledgeable individuals and means for information acquisition and dissemination in more populous places, we expect the positive impact of large cities on firm productivity to be especially profound in team-based organizational structures. Indeed, teams in such places could benefit from a larger pool of skilled workers and/or from better information that each team member may possess, acquire, and disseminate among other group members. Therefore, the environment of large cities may have a compounding effect on the performance of firms operating under a collective decision-making.

Mutual fund data is ideal for analyzing links between urban geography, organizational structure, and productivity, since fund industry provides the most comprehensive source of occupational data with both team-managed and single-managed funds, follows the task of maximizing performance, and has presence in various locations around the country. Our fund and fund manager data comes from Morningstar Direct covering the 1992-2010 period. We account for actively managed U.S. domestic equity funds with four investment objectives: aggressive growth, growth, growth & income, and equity income. Our set of U.S. metropolitan areas includes 91 cities, out of which 80 have team-managed fund operations, 76 single-managed ones, and 65 localities record both team-managed and single-managed funds. The sub-sample of the most populous metro areas consists of the top seven areas based on the 2000 U.S. Census data. It includes Boston, Chicago, Los Angeles, New York, Philadelphia, San Francisco, and Washington, D.C.

We first examine fund performance with different managerial structures in relation to metro area population while controlling for two other relevant demographic variables: per capita income and education level (proportion of people with university degree) in each metro area. We find that team-managed funds post a positive and significant relation between area population and fund returns adjusted for investment objective. There is limited or no evidence on the relation of funds returns to other demographic variables.

Then, we move to our main tests and show that team-managed funds outperform singlemanaged funds in the largest metro areas but not in other places. This result is present irrespective of the fund performance evaluation metrics, robust to the inclusion of fund and manager controls, and exists in both the first and second parts of our sample period. In economic terms, the average contribution of team management to fund performance in large cities is at least 50 basis points (bps) per year higher than that of single-managed funds or team-managed funds in smaller metro areas. Furthermore, we obtain similar results when splitting the sample based on the median education level in urban areas: team-managed funds in more educated cities add value to fund performance, but those in less educated places do not. Yet, the primary contributors to this result are still large metro areas of Atlanta, Denver, and Minneapolis, with well-developed fund industry that replace Chicago, Philadelphia, and Los Angeles. Across investment objectives, we observe the lowest contribution of team-management to funds returns for aggressive growth and growth funds in small metro areas.

In addition, we look how the experience working in a team affects fund performance. We find that in large metro areas the outperformance of team-managed funds is driven by those funds whose managers have longer fund tenure. Their risk-adjusted returns can exceed 53 bps per year those with shorter tenure in larger metro areas; they also exceed those of similar team-managed funds in smaller localities. Moreover, we obtain a similar, yet stronger picture when we account not only for the same number of managers within the team but also for the same individuals comprising the team after its formation. The difference in risk-adjusted and characteristic-adjusted returns between fund teams located in the largest areas that are formed four or more years ago and those formed three or less while marinating the same managers, is between 44 bps and 70 bps per year.

Finally, we also consider portfolio holdings data and evaluate differences between teammanaged and single-managed funds across metro areas in terms of the average proportion of holdings and active share investing. We observe that team-managed funds in large metro areas hold significantly more concentrated portfolios and active share than their peers in smaller cities. These results are consistent with Kacperczyk, Sialm, and Zheng (2004) and Cremers and Petajisto (2009) who show that funds holding more concentrated portfolios or those having a high active share achieve better performance.

Thus, we make two contributions to the literature. First, our results provide a novel piece of evidence in support of the learning theory of urban agglomeration of Jacobs (1969), Lucas (1988), Glaeser, Kallal, Scheinkman, and Shleifer (1992), and Glaeser (1999). Had initial skills of fund managers in larger metro areas were the dominant factor, one would have expected to see the outperformance of team-managed funds across all teams, irrespective of the team formation time. Yet, the teams of mutual fund managers in the largest metro areas are unable to outperform single-managed funds until after their members gain some experience of working together. Second, our findings also add to the recent studies that are able to detect superior performance of team-managed organizations in an empirical setting (e.g., Hamilton, Nickerson, and Owan, 2003). Our contribution here is in showing that the benefits of more information flows and diversity of opinions in teams are more profound in an environment which itself is more conductive to information generation and dissemination.

The rest of the article is organized as follows. Section 2 describes the fund data as well as U.S. demographic variables. Section 3 gives preliminary evidence of the differences in teammanaged fund performance across U.S. metro areas. Section 4 presents the main empirical findings of our paper. Section 5 examines the relation between team performance and team working experience as well as the amount of time since team formation. Section 6 repeats our main empirical estimations with three alternative fund performance measures. Section 7 concludes.

2. Data

2.1. Fund and Manager Characteristics

Our mutual fund data comes from Morningstar Direct and covers a period from 1992 to 2010. At present, it provides the most accurate data source for U.S. mutual funds (see comparison with CRSP and Morningstar Principia datasets in Patel and Sarkissian, 2016). We consider all actively managed U.S. diversified domestic equity funds that fall into the four investment objectives: aggressive growth, growth, growth & income, and equity income. All index and sector funds are excluded from our analysis. Morningstar Direct reports all data at the fund share class level, including the names of the fund managers. To avoid multiple counting of fund management information from different share classes of the same fund, we aggregate share class level observations to one fund level observation.

For each fund we obtain information on the following characteristics: fund size, measured by the total net assets under management of the fund at the end of calendar year; fund age, defined as the difference between the fund's inception year and the current year; expenses, measured by the annual net expense ratio of the fund; turnover, measured by the turnover ratio of the fund; fund family size, measured by the total net assets under management of the fund complex to which the fund belongs at the end of calendar year; fund return volatility, measured by standard deviation of raw net returns of funds over the past year. We also include net fund flows, defined as the net growth in the total net assets of funds, as a percentage of their total net assets, adjusted for prior year returns. To minimize the effect of outliers on our analysis, expense ratios, turnover, and annual fund flow are winsorized at 1% and 99% levels.

We classify a fund as single- or team-managed based on the number of fund managers with the fund at the end of calendar year. When only one fund manager is named at the end of calendar year, we classify that fund as sole-managed for that year. Similarly, when two or more fund managers are named with the fund, we classify the fund as team-managed. We remove all fund-years which have missing or anonymous fund manager names or tenure dates from our sample.⁴ Our final sample covers 3,008 unique funds with 29,166 manager-fund-year observations.

Chevalier and Ellison (1999) show the importance of managerial characteristics for fund returns. Moreover, since our goal is to see whether group decision-making has across contribution to fund performance in various locations, it is necessary to control for other possible manager-specific characteristics. Therefore, following Chevalier and Ellison (1999), Barber and Odean (2001), and others, we create three manager characteristics variables: tenure, average MBA and female variables.⁵ We define the manager tenure as the difference between the year when a fund manager started as a portfolio manager for a given fund and the current year. We define the MBA variable as the proportion of fund managers in a team with an MBA degree. Likewise, we define the Female variable as the proportion of female fund managers. To create a managerial team characteristic, we simply assume equal contribution of each team member as in Patel and Sarkissian (2016). Therefore, manager tenure of the team is simply the equally-weighted average of manager tenure of each fund manager in the team, respectively.

2.2. Demographic variables

Our main demographic variable is the total population of U.S. metropolitan areas as defined by the 2000 U.S. Census. Christoffersen and Sarkissian (2009) show that fund managers located in financial centers earn higher returns than their peers located in smaller towns. To account for this effect, we obtain the location information of fund advisors. It is important note that our location variable differs from the previous studies. Instead of using the headquarter

⁴ The proportion of blank or anonymous entries for fund manager information in our initial data sample is only 7%. This stark difference with the percentage of anonymous funds reported in Massa, Reuter, and Zitzewitz (2010), which was reaching 18% in some years is due to the fact that Morningstar Direct has filled in names of managers for almost all funds (retroactively) after 2006.

⁵ Unlike Chevalier and Ellison (1999), we do not use manager age variable for three reasons. First, it is based on indirect information on undergraduate degree year information. Second, it is available for a much smaller set of funds. Third, it is sizably correlated with manager tenure variable. Also, we do not consider the average SAT scores of managers, since this data again greatly reduces our sample. In unreported results, we find very little impact of the SAT score inclusion on our estimations.

location of the fund company or fund sponsor, we use the headquarter location of the fund advisor company. For majority of funds, the fund advisor and the fund sponsor (the company that offers the mutual fund to public) might be the same company (Chen, Hong, and Kubik, 2013). But for few funds they might be different because these funds choose to outsource their portfolio management to third-party fund advisor companies. By choosing the fund advisor location, we make analysis immune to the possibility of any bias due to third-party fund management outsourcing.

There are 91 metro areas with the population between 66 thousand and 21 million people that have fund advisors in our sample. The median metro area for all our fund-year observations is Boston with the total population of 6 million people. Therefore, the largest U.S. urban agglomerations in descending order based on the 2000 U.S. Census are New York, Los Angeles, Chicago, Washington, D.C., Philadelphia, San Francisco, and Boston. These agglomerations, with the exception of Washington, D.C., are the same as in Hong, Kubik, and Stein (2005) and Christoffersen and Sarkissian (2009): in those studies they constitute the largest U.S. mutual fund or financial centers. If the fund advisor company is headquartered within a 50-mile radius of any of these seven cities, we classify such fund as located in the large metro area.

We also collect information from the same Census on two other demographic variables for each metro area: the per capita income and the proportion of people with a university degree. The per capita income ranges between \$15,000 (Fresno, CA) and \$31,000 (Naples, FL), while the proportion of bachelor degree holders – between 14.6% (Wheeling, WV) and 40.6% (Madison, WI). These two variables can be considered as proxies for more skilled on average metro areas. If initial set of skills is important for better collective decision-making, one could expect to see a positive relation between team-managed funds and one or both of these variables.

2.3. Fund Performance Measures

In our main estimations, we use three different performance metrics: a four-factor alpha, $\alpha(4F)$, using Carhart (1997) model, a five-factor alpha, $\alpha(5F)$, coming from the addition of the

Pastor and Stambaugh (2003) liquidity factor to the Carhart (1997) model, and the characteristic selectivity measure, CS, of Daniel, Grinblatt, Titman, and Wermers (1997). That is, we estimate each fund's $\alpha(4F)$, and $\alpha(5F)$, using the following two equations:

$$r_{i,t} = \alpha_i (4F) + \beta_i r_{m,t} + s_i SMB_t + h_i HML_t + m_i UMD_t + e_{i,t}$$

$$\tag{1}$$

and

$$r_{i,t} = \alpha_i(5F) + \beta_i r_{m,t} + s_i SMB_t + h_i HML_t + m_i UMD_t + l_i LIQ_t + e_{i,t}, \qquad (2)$$

respectively, where $r_{i,t}$ is the monthly gross fund return less the risk-free rate (the one-month U.S. T-bill rate), $r_{m,t}$ is the monthly U.S. excess market return (the return on the CRSP value-weighted NYSE/AMEX/Nasdaq composite index less the one-month U.S. T-bill rate), while α_i is the risk-adjusted return. SMB, HML, and UMD are returns on the size, book-to-market, and momentum portfolios, respectively.⁶ LIQ is the liquidity factor of Pastor-Stambaugh (2003). The CS measure for fund *i* is computed as follows:

$$CS_{i,t} = \sum_{j=1}^{N} w_{j,t-1} \Big(R_{j,t} - R_t^{b_{j,t-1}} \Big),$$
(3)

where $w_{j,t-1}$ is the portfolio weight on stock *j* at the end of month *t*-1, $R_{j,t}$ is the stock return in month *t*. $R_t^{b_{j,t-1}}$ is the characteristic-based benchmark portfolio return in month *t* that is matched to stock *j* in month *t*-1.⁷ The matching is done on a basis of size, book-to-market-ratio, and momentum.

The managerial structure of funds changes annually. Therefore, we remove all fund-years that have less than twelve monthly return observations and estimate the fund performance metrics using their prior twelve monthly returns. This procedure is immune to the reverse survivorship bias of Linnainmaa (2013) and does not introduce bias by incorrectly attributing

⁶ This data is from Ken French's site: http://mba.tuck.dartmouth.edu/pages/faculty/ ken.french/data_library.html.

⁷ Characteristic-based benchmarks are from Russ Wermer's site: http://alex2.umd.edu/wermers.

fund performance to a certain type management structure. To reduce the influence of outliers coming from our short fund performance estimation windows, we trim $\alpha(4F)$, and $\alpha(5F)$, which, unlike the CS measure, are regression-based, at the top and bottom 1% of the distribution.⁸ We also define objective-adjusted return, OAR, as the difference between the average monthly gross return of a fund in the year minus the mean fund returns across all funds for a given fund investment objective and year.

2.4. Summary Statistics

First, Figure 1 shows the distribution of the average number of fund managers across 65 U.S. metro areas that have both team-managed and single-managed funds. The horizontal axis shows the metro population in millions. The leftmost observation belongs to Casper, Wyoming, while rightmost observation to New York. The team size here is the average number of fund managers in each location. The fund team size in excess of four managers is assumed to consist of four members. Bars closer to one and zero indicate localities with predominance of small team and single-managed funds. We can observe that team-managed funds are common across the entire sample of metro areas without any visible relation to population size. In fact, the average number of fund managers is 2.02 both for the smallest 33 areas and the largest 32 areas. However, it is not clear whether having multiple portfolio managers is beneficial to fund performance in all locations.

Table 1 provides the detailed summary statistics. Panel A shows the distribution of funds and fund managerial structure across each of the seven largest U.S. metro areas and other locations. Large population metro areas (LPAs) include 1,802 distinct funds; small population metro areas (SPAs) – 1,206 funds. New York has the largest number of funds – 718, while Washington, D.C. the smallest – only 32. Across funds with manager team sizes, we can note that single-managed funds constitute the majority in most of the locations except Los Angeles

⁸ Smoothing of twelve-month fund returns is common in the literature (e.g., Cassar and Gerakos, 2011; Aiken, Clifford, and Ellis, 2013; Chen, Hong, Jiang, and Kubik, 2013). Our test results based on winsorized risk-adjusted returns, which are similar to those in this study, are available on request.

and Washington, D.C., where the most common form of managerial structure is two-manager teams. Boston appears to have the largest fraction of single-managed funds among all other largest metro areas: it exceeds 50% of all fund observations.

Panel B of Table 1 shows basic statistics, such as the mean, standard deviation, median, and the difference in the means test, for fund and manager characteristics in large and small metro areas. Consistent with Christoffersen and Sarkissian (2009), we find that objective-adjusted returns (OARs) in larger metro areas are on average significantly higher than in smaller locations. Large metro area funds also tend to be larger and older, belong to larger fund families, have much higher annual turnover, but lower expense ratio than small metro area funds. There are significant differences in managerial characteristics across locations as well. The manager tenure with the same fund in large metro areas is significantly lower than that in smaller metro areas, although the median of three years is the same across all locations. Also, on average, funds in larger cities have more managers with an MBA degree and more females.

3. Preliminary Evidence

In the fund management industry, skills, knowledge, as well as networking ability of each team member can be of great importance to fund performance. That is, if teams in the financial industry are able to achieve diversification of style and judgment, as argued by Barry and Starks (1984), and Sah and Stiglitz (1986, 1991), then the value of having a team must be larger when each individual has a higher potential to enhance the overall knowledge and resource base of the group. Numerous studies have shown that those conditions are more readily available in larger urban agglomerations. This could be due to a more vibrant pool of skilled workers, who find appropriate jobs easier in larger metro areas, as argued by Helsley and Strange (1990, 1991). Alternatively, it could be due to an easier knowledge transfer, faster and more diverse business and social connections, and potential access to private information found in larger cities, as

argued by Jacobs (1969), Glaeser (1999), and Christoffersen and Sarkissian (2009). These two explanations are not mutually exclusive, and our task is to provide sufficient evidence for one or both potential reasons.

In Table 2 we regress objective-adjusted returns of team-managed and single-managed funds on the metro area population as well as on the two other demographic variables: average per capita income (in thousand U.S. dollars) and education level (the percent of population with a university degree) in each metropolitan area. In regressions, the population data are transformed logarithmically. All estimations are based on the Huber-White robust standard errors. Panel A uses fund returns across all metro areas with team-managed and single-managed fund return observations. There are 80 locations with team-managed fund returns and 76 singlemanaged. In Panel B we include only those metro areas that have at least ten fund return observations. The main conclusion from these two panels is that, indeed, there is a positive and significant relation between fund performance and metro area population size. Yet, there is no any relation between fund returns and the average income. There is also some indication that education level of the metro area may be conductive to team-managed fund performance. The performance of single-managed funds shows no relation to any of the three demographic variables. Thus, Table 2 results highlight a distinct impact of metro areas with different population sizes and, to some extent, average education level to the performance of teammanaged funds.

In general, larger cities are associated with larger educational opportunities. Indeed, Table 2 shows a positive relation between the log of population and percent of university degree holders in 91 U.S. metro areas that have domestic equity mutual fund data in our 1992-2010 sample period. This relation is significant at the 5% level. In our data, the median percent of university degree holders across 91 cities is 30.5%. We denote cities with education level exceeding (not exceeding) the sample median as high (low) education areas, HEA (LEA). Among our sample of seven large metro areas, four, namely: Boston, New York, Washington, D.C., and San Francisco also have above median education level. Among cities with high education level but outside our set of the largest metro areas, Atlanta, Denver, and Minneapolis, also provide substantial number of fund return observations. In fact, these cities also belong to the largest urban areas in the United States – more precisely, the top 20% in our sample of 91 areas with mutual fund companies. Therefore, in some of the follow-up tests, we perform estimation on metro area sub-sample splits by both population size and education level.

In Table 3 provides the first formal tests of the differences in team-managed fund performance across locations using our three performance measures, $\alpha(4F)$, $\alpha(5F)$, and CS. It shows the means and standard deviations of these metrics and the difference in means test between team-managed and single-managed funds returns for each location set. Panels A and B report these statistics for large and small metro areas, respectively. We can see that the differences in $\alpha(4F)$ and $\alpha(5F)$ between team-managed and single-managed funds are significant in large metro areas. In economic terms, in large metro areas the risk-adjusted returns of teammanaged funds are 26 bps (0.0217×12 or 0.220×12) per year higher than similar returns of single-managed funds. However, we observe no differences in $\alpha(4F)$ and $\alpha(5F)$ between teammanaged funds in small cities. In addition, while the difference in CS measure is negligible in large cities, team-managed funds significantly underperform their single-managed counterparts in smaller metro areas.

Panels C and D in Table 3 report our performance statistics for high and low education metro areas, respectively. The overall picture is similar to that in Panels A and B. Again, the differences in $\alpha(4F)$ and $\alpha(5F)$ between team-managed and single-managed funds are economically and statistically significant in large metro areas but not small ones. The magnitude of the two risk-adjusted returns of team-managed funds in high education areas is 37-43 bps per year higher than that for similar returns of single-managed funds. The difference in CS measure is insignificant in Panel C, yet is negative and highly significant in Panel D, indicating underperformance of team-managed funds in low education metro areas. Thus, Table 3 shows directly, albeit without accounting for fund and manager characteristics, that team management

in the fund industry is associated with better performance only in large and more educated metro areas but not in smaller and low education cities.

4. Main Empirical Tests

We now move to examining the relation between fund performance and team management in large and small metro areas while controlling for the full set of lagged fund and managerial characteristics.⁹ The regression model is as follows:

$$Perf_{it} = c_0 + c_1 Team_{it-1} + \delta_1 Fund _Controls_{it-1} + \delta_2 Mgr_Controls_{it-1} + \delta_3 FE_{it} + e_{it}, \qquad (4)$$

where $Perf_{i,t}$ is one of our three measures of fund performance, Team is the dummy for teammanager funds, Fund_Controls and Mgr_Controls are fund- and manager-specific characteristics, respectively. The fixed effects, FE, include the year times the investment objective effects, fund family fixed effects, as well as the metro area dummies.¹⁰ The fund characteristics are lagged by one period to exclude their contemporaneous influence on fund alphas. We also use an alternative specification to Eq. (4) to estimate directly the impact of the team management between large and small metro areas and between metro areas with high and low average education level. These regressions are:

$$Perf_{i,t} = c_0 + c_1 Team_{i,t-1} \times LPA_i + c_2 Team_{i,t-1} \times SPA_i + \delta(Controls and FE) + e_{i,t},$$
(5)

and

$$Perf_{i,t} = c_0 + c_1 Team_{i,t-1} \times HEA_i + c_2 Team_{i,t-1} \times LEA_i + \delta(Controls and FE) + e_{i,t},$$
(6)

⁹ Lagging the Team dummy precludes the look-ahead bias in our estimations.

¹⁰ Metropolitan area fixed effects are included given the evidence that mutual fund returns are different across geographic locations (see Coval and Moskowitz, 2001; Christoffersen and Sarkissian, 2009).

where Team×LPA and Team×SPA are the interactive variables of the team dummy with dummies for large and small population metro areas, while Team×HEA and Team×LEA are those for high and low education areas. In models (5) and (6) the sets of control variables and fixed effects are the same as in model (4).

4.1. Teams, location, and fund performance

Table 4 shows the estimates from panel regressions of $\alpha(4F)$, $\alpha(5F)$, and CS on Team dummy and other controls using Eq. (4) across locations based on metro area population size. It also reports the p-values of coefficients, the number of observations, and the R-squared. The standard errors are clustered by fund and year. Panel A shows the full sample period results. The first three columns show the regression estimates for the largest seven metro areas. The Team coefficient is positive and highly significant in all three specifications. In economic terms, in large metro areas team-managed funds outperform their single-managed counterparts by 51-67 bps per year based on fund alphas and 130 bps per year based on the characteristic selectivity. In the last thee columns of Panel A, which show regression estimates for smaller metro areas, we can see that across all specifications the coefficient on Team is statistically zero and economically very small as well. This implies that on average teams add no gains to performance for funds located outside the largest seven U.S. metro areas. Moreover, the last two lines in the panel give the difference in the Team coefficient across metro areas, Team (LPA-SPA), and the corresponding p-value of the F-test for the equality of estimates. Here, we see that the riskadjusted return difference between team-managed funds located in large metro areas and those in small metro is again economically and statistically significant at the 5% level. The difference in the corresponding CS measures is significant at the 10% level.

It is worthwhile to note that out of the eleven independent variables in regression model (4), the impact of the Team dummy is the most different one between large and small metro areas. All other variables have very similar influence on three fund performance metrics,

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irrespective of the location of fund advisors. We also note that the overall fit of the regression with CS as a dependent variable is much lower than that with risk-adjusted returns.

Panel B of Table 4 shows the results for the 1992-2004 and 2005-2010 sub-periods. This unequal time period split yields comparable number of fund-year observations in both sub-samples. In this panel, we use estimations based on Eq. (5). For conciseness, we do not report the estimates on fund and manager controls. Our results are very similar to those over the whole sample period in Panel A. As before, in large metro areas, irrespective of the performance measure and the time period, the Team coefficient is always positive and significant both economically and statistically, except for the CS measure prior to 2005. Note that the reduction in statistical significance in this case is largely due to the smaller sample size. Yet, again similar to the full sample results, there is no any evidence of performance benefits among team-managed funds in small metro areas.

Recall that in Table 3 we observe that the impact of team management on fund returns varies also across cities' education levels. Therefore, in Table 5 we conduct tests, similar to those in Table 4, but splitting the sample by high and low education metro areas. In this Table, we report only the coefficients on the variables involving the Team dummy, although all controls are similar to those in Table 4. Panel A reports the full sample period results. Consistent with Table 3 findings, team-managed funds show significant outperformance only in high education areas. The economic impact of team management for risk-adjusted returns is 50-68 bps, which is remarkably similar to the corresponding estimates in Panel A of Table 4. The CS measure is significant economically, but not statistically. In contrast, the Team dummy is not only statistically insignificant for low education level areas but also takes economically non-trivial negative values in these locations for all three performance metrics. The last two lines in the panel, show that team-managed funds in high education areas outperform their peers in low education areas by as much as 68-86 bps per year based on $\alpha(4F)$ and $\alpha(5F)$.

Panel B of Table 5 shows the regression results using Eq. (6) for the 1992-2004 and 2005-2010 sub-periods. The format of this panel is identical to that of Panel B of Table 4. The

outcome of estimations is also similar to previous findings. In both sub-periods, the coefficient on the interactive term, Team×HEA, is positive and significant for all three measures of fund performance (except for the CS measure statistically), while that on Team×LEA is small and insignificant. This again reflects the fact that teams provide no additional value to fund performance outside a very select group of cities with highly educated average workforce. Since these metro areas are still large urban agglomerations, all our subsequent sample splits are done using only metro area population size.

While our finding on the outperformance of team-managed funds in large or highly educated metro areas withstands sub-period tests, it is also necessary to observe whether this outcome is concentrated in certain types of funds. One can expect that the significance of team management should be positively related to the ability of funds collect information on hard-tovalue assets. Since teams indeed add to fund returns in larger cities, team-managed funds located in these metro areas should be in a relatively better positioned in dealing with growth stocks than those with steady dividend income.

To examine this, in Table 6 we show tests on how managerial structure impacts funds returns in different locations across two funds groups composed from four original funds investment objectives. The first group, which we call Growth funds, contains aggressive growth and growth funds; the second group, which we call Income funds, contains growth & income and income funds. All the econometric specifications of regressions are the same as in Table 4, Panel B, that is are based on Eq. (5). The table also reports the difference in the Team coefficient across the two metro area groups and the corresponding p-value of the F-test for the equality of estimates. We can see that team management is uniformly more profound in large metro areas for Growth funds only. The coefficients on Team×LPA is positive and statistically significant for all three performance metrics (except statistically for the CS measure). Its economic importance for risk-adjusted returns exceeds 48 bps per year. The coefficient on Team×SPA is effectively zero. In contrast, for the Income group of funds, the team-managed funds in both large and small metro areas show comparable performance relative to the corresponding single-managed funds.

Thus, Tables 4-6 provide comprehensive statistical evidence that the benefits of team management are widely different across urban agglomerations with different population size. The gain in risk-adjusted returns from the team-based approach in portfolio management is significantly higher among those funds, whose advisors are located in larger and more educated metro areas. This result holds in both the earlier and the later sample periods and is manifests itself more profoundly among funds with growth investment objectives. Our findings support Sharpe (1981) arguments and provide novel evidence that group decision-making is more beneficial in such environments where their members are more likely to acquire and use knowledge, skills, and establish business connections. In the finance industry in general and fund industry in particular, this becomes more achievable in larger and more affluent cities.

4.2. Alternative Performance Measures

In this section, we repeat our main estimations using alternative performance evaluation models. We consider three new metrics. The first one is the Fama-French three-factor alpha $\alpha(3F)$, computed from

$$r_{i,t} = \alpha_i (3F) + \beta_i r_{m,t} + s_i SMB_t + h_i HML_t + e_{i,t},$$
(7)

where, as before, $r_{i,t}$ is the monthly gross fund return less the risk-free rate.

The second alternative is the conditional Carhart (1997) alpha, $\alpha(4C)$. We use a specification, where market beta is modeled as a linear function of instruments, as in Ferson and Schadt (1996), namely:

$$r_{i,t} = \alpha(4C)_i + \beta_i r_{m,t} + s_i SMB_t + h_i HML_t + m_i UMD_t + b_i^{Tbill} r_{m,t} Z_{t-1}^{Tbill} + b_i^{Term} r_{m,t} Z_{t-1}^{Term} + e_{i,t}.$$
(8)

In Eq. (8), Z_{t-1}^{Tbill} and Z_{t-1}^{Term} are the two lagged (demeaned) public information variables: the onemonth U.S. Treasury bill rate (T-bill) and the term-structure spread (Term), defined as the difference in yields on the 10-year U.S. government bond and three-month U.S. T-bill. We again trim the top and bottom 1% of distribution of both risk-adjusted returns.

The third alternative is the return gap, RG, of Kacperczyk, Sialm, and Zheng (2008). Return gap is which is defined as the difference between the reported fund return and the return of a portfolio that invests in the previously disclosed holdings net of expenses. Table 7 shows the estimation results with alternative performance measures. In the first three columns of the table the estimations are based on Eq. (5), where team-management impact is compared between large and small population size metro areas. In the last three columns of the table the estimations are based on Eq. (6), where team-management impact is compared between high and low education metro areas. Our results are very similar to previous findings for $\alpha(3F)$ and $\alpha(4C)$. We see highly positive and significant coefficients on Team×LPA and Team×HEA but not Team×SPA and Team×LEA for these performance measures. Moreover, a similar pattern is observed for the return gap in column 6. Only in column 3 for the RG measure the coefficient on interactive Team variable shows no difference between locations. Thus, we can state that the evidence on outperformance of team-managed funds in larger and more educated cities relative to their counterparts in smaller places depends very little on the chosen performance evaluation metrics.

4.3. Team Formation and De-Formation

There are two potential issues with our previous tests. First, we do not differentiate between teams of different sizes. Hamilton, Nickerson, and Owan (2003) and Patel and Sarkissian (2016) show that team size has a non-linear relation to firm performance. Second, in those tests we do not maintain the same fund managers within the team. It is clear that for a given team, there can be differences in both collaboration gains and coordination costs if over time the members comprising the team are different people or are the same individuals. Our next tests address these points by comparing the returns of team-managed and single-managed funds over time with their own past performance. Therefore, we now focus our attention only on those team-managed and single-managed funds that change their managerial structure within our 19

year sample period, and call these instances as team-formation and team de-formation. All funds with a constant managerial structure throughout the sample period are dropped. This condition significantly lowers the overall number of available observations.

We define the team formation year to be τ if fund *i* is team-managed at times *t*, *t*-1,..., *t*- τ , but was single-managed at time *t*- τ -1. Similarly, the team de-formation year is τ if fund *i* is single-managed at times *t*, *t*-1,..., *t*- τ , but was team-managed at time *t*- τ -1. Figure 3 shows the stability of the managerial structure is from the team formation and de-formation years. Plot A depicts the average percent of the same managers in a team after the team formation for the given fund. The first year of team formation is shown with 100%. However, by the second year, about 25% of team members are replaced. The subsequent changes in team members are not so drastic, but the overall number of the same managers tends to decrease slowly. Plot B gives the average percent of the same single manager after team de-formation with the given fund. Here, the initial change in the manager at year two is constitutes only about 10%. In subsequent years, the change of the single manager becomes an even rarer event.

Table 8 examines the relation between the formation and de-formation of teammanagement and fund performance across metro areas of different population sizes. New (old) team formation denotes funds that became team-managed up to three (four or more) years ago, while new (old) team de-formation denotes funds that became single-managed up to three (four or more) years ago. It also shows the p-value for the difference test between old and new team formation/de-formation, Diff (Old – New). Panel A presents the results for the same number of managers. We can see that the average risk-adjusted returns for teams in large metro areas that were formed more than four years ago are significantly higher than those for teams formed more recently for all three performance metrics. The average difference is about 53 bps for $\alpha(4F)$ and $\alpha(5F)$ and 40 bps for the CS measure. There are no significant differences between singlemanaged funds which are formed recently or late.

Panel B of Table 8 gives the test results for the same portfolio managers within each team. The number of observations in this panel is smaller than that in Panel A since substantial

changes occur over time among team members even within teams of the same size. In spite of this difference, the overall pattern of team-managed and single-managed funds performance is similar to that in Panel A. Again, we can see that $\alpha(4F)$, $\alpha(5F)$, and CS for teams in large metro areas formed more than four years ago are much larger than the corresponding returns for teams formed in the last three years. Moreover, the magnitude of these differences is even higher than before, reaching 65 bps for $\alpha(4F)$, 70 bps for $\alpha(5F)$, and 44 bps for CS. Thus, Table 6 results have very important implication to our understanding on how teams improve on fund performance in larger cities. It appears that it is not the inner initial set of skills of group members in larger cities that plays a crucial role there, as suggested by Helsley and Strange (1990, 1991). Rather, it is the wider set of opportunities which large cities offer for knowledge collection and production. It is not surprising then that these unique knowledge dissemination and learning possibilities in large metro areas become revealed in team work only over time, which is consistent with Jacobs (1969), Lucas (1988), Glaeser (1999), and others.

5. Portfolio Holdings

Numerous studies document certain portfolio holding patterns among outperforming mutual funds. For example, Kacperczyk, Sialm, and Zheng (2004) find that funds with more concentrated investments post higher returns than those investing in more diversified portfolios. It is known that equity mutual funds hold only a small fraction of their holdings in cash, T-bills, and other fixed-income securities. Therefore, funds with lower number of holdings are expected to post higher average returns. Given wide performance differences among team-managed funds between large and small metro areas, one should expect to similar patterns in their portfolio holdings as well.

Table 9 reports the relation of management structure on fund holdings across the two location sets. The econometric framework is similar to Eq. (5), except that the dependent

variable now is the log of portfolio holdings for each fund-year. As before, all regression specifications include time by investment objective, fund family, and metro area fixed effects, and the standard errors are clustered by fund and year. The first three regressions use the full sample period and different sets of control variables. We can see that, irrespective of controls, the coefficients on both interactive Team terms are positive and significant. This shows as in Patel and Sarkissian (2016) that team-managed funds in general hold more securities. What is much more important in this study, however, is that the magnitude as well as statistical significance of the coefficient on Team×LPA is consistently *lower* than that on Team×SPA. The F-test results at the bottom of the table highlight the difference in these coefficients. The sample period slit in the last two columns of the table presents the same pattern. The only exception is that, due to the significantly smaller sample size, the F-tests in these cases do not record statistical significance in the difference. This means that fund manager teams in large cities hold relatively smaller number of securities than comparative teams managing similar funds in smaller metro areas. After analyzing the composition of aggregate holdings, we observe that team-managed funds in large cities invest significantly more in stocks and significantly less in bonds and cash than team-managed funds in smaller metro areas.¹¹

With similar reasoning, we should also expect differences in team-managed funds across locations in terms of active share holdings. Cremers and Petajisto (2009) show that active share holdings, which represents the proportion of holdings that differs from the benchmark index holdings, is higher among better performing funds. Therefore, active share among team-managed funds in larger should be higher than that among similar funds in smaller areas. Table 10 illustrates the test results. In this table the regression framework is again based on Eq. (5) but fund performance is substituted with active share for each fund-year.¹² The structure of the table is similar to that in Table 9. The first three columns show the estimation results for various sets of control variable, while the last two perform sub-period estimation. We find that the coefficient

¹¹ These results are available on request.

¹² The active share data is kindly provided by Martijn Cremers.

on the interactive Team dummy in large metro areas is small and insignificant in all five columns of the table, implying that active share among team-managed and single-managed funds in large cities is about the same. Yet, the coefficient on Team×SPA is negative and significant also across all estimations, and the last two lines in the table indicate that the difference with Team×LPA is significant too. This suggests that team-managed funds in smaller metro areas have lower active share with respect to both team-managed funds in large cities as well as practically all singlemanaged funds on average.

Thus, Tables 9-10 show that, consistent with outperformance results among teammanaged funds in large metro areas documented in earlier sections, their portfolio holdings also show quite distinct patterns. These funds have a much higher propensity of holding more concentrated portfolios as well as higher active shares than team-managed funds in smaller metro areas.

6. Conclusions

In this paper, we connect urban economics and organization structure. A large number of studies in both strands of literature suggest that better information processing environment can enhance firm productivity. The urban economics literature relates the areas with larger population as more conductive to learning and knowledge spillover. The literature on organization structure suggests that teams generate better solutions from diversified knowledge and skills of their members. As a result, it becomes quite suggestive that team management must lead to higher gains in large metropolitan areas.

Indeed, using U.S. equity mutual fund and demographic data, we find that the benefits of team-based managerial approach are drastically different across locations. Team-managed funds outperform single-managed funds in larger metro areas but not in small ones. This outcome is robust across various fund performance metrics, sub-sample periods, and the inclusion of fund

and manager characteristics controls. Moreover, we show that superior performance of teammanaged funds in large metro areas in present only among portfolio managers with substantial team-work experience. There are no differences in risk-adjusted and characteristic-adjusted returns among recently formed teams of funds managers in relation to single managers, irrespective of fund location, implying that the initial skills of group members are not very important. We also observe differences in security holding of team-managed funds between large and small cities. Team-managed funds in large agglomerations hold more concentrated portfolios and have higher active share. Therefore, we interpret our results as evidence of the positive externality of larger cities for team-based organizational structures through more accessible information sources, larger social and business networks, and better learning opportunities.

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Table 1 Summary statistics

Panel A: Distribution of funds

		Manager team size					
	Obs	1	2	3	4+	# Funds	
Large metro areas	16,689	7,093	4,394	2,022	2,394	1,802	
Boston	3,531	1,893	831	403	264	381	
Chicago	1,748	658	492	237	278	208	
Los Angeles	1,029	306	316	102	252	114	
New York	6,743	2,751	1,933	836	906	718	
Philadelphia	1,555	550	369	200	349	167	
San Francisco	1,790	855	359	223	288	182	
Washington, D.C.	293	80	94	21	57	32	
Small metro areas	12,477	4,508	3,019	1,418	1,805	1,206	
Total	29,166	11,601	7,413	3,440	4,199	3,008	

Panel B: Fund and manager characteristics

	Metro area	Obs	Mean	S.D.	Median	L - S
OAR (%/m)	L	16,659	0.038	1.192	0.026	0.032 ^{**}
	S	12,197	0.006	1.423	0.006	(0.036)
TNA (bln \$)	L	17,977	1.227	5.623	0.149	0.500 ^{***}
	S	13,124	0.727	2.212	0.114	(0.000)
Family TNA (bln \$)	L	19,123	29.41	77.54	6.446	13.51 ^{***}
	S	14,064	15.90	27.01	2.438	(0.000)
Fund Age	L	18,669	10.810	13.15	7.000	0.895^{***}
	S	13,649	9.915	11.28	7.000	(0.000)
Turnover	L	16,305	0.965	1.120	0.720	0.102 ^{***}
	S	11,942	0.863	1.496	0.620	(0.000)
Expenses (%/y)	L	15,943	1.282	0.563	1.233	-0.067 ^{***}
	S	11,664	1.349	1.212	1.225	(0.000)
Manager Tenure	L	18,617	4.017	4.004	3.000	-0.157 ^{***}
	S	13,600	4.174	4.216	3.000	(0.001)
MBA	L	18,617	0.522	0.408	1.000	0.068^{***}
	S	13,615	0.454	0.407	1.000	(0.000)
Female	L S	18,598 13,582	0.096 0.076	0.237 0.206	$0.000 \\ 0.000$	0.020 ^{***} (0.000)

Table 1 (continued)

This table gives the summary statistics of domestic equity mutual funds in the United States from 1992 to 2010 for large and small metro areas. The fund data comes from Morningstar Direct. The metro areas are classified based on the 2000 U.S. Census. Panel A shows distribution of funds across the largest metro areas and fund manager team size. Panel B gives fund and manager characteristics between the two locations and provides their difference test. OAR (%, per month) is the investment objective adjusted fund return, which is the difference between the average monthly net fund return for fund i in year t and the average monthly fund return of all funds in the matched investment objective in year t. TNA (\$, billions) is the total net asset under management of fund i in year t. Family TNA (\$, billions) is the total net asset under management of fund i in year t. Fund Age (years) is the difference between fund i's inception year and the current year t. Turnover is the minimum of aggregated sales or aggregated purchases of securities of the year divided by the average 12-month total net assets of the fund. Expenses (%) is the annual total expense ratio of the fund i in year t. Manager Tenure (years) is the number of years the fund manager remains with the fund i at time t. MBA is defined as a proportion of managers in a fund with a MBA degree. Female is defined as a proportion of female managers in a fund. In case of teams, the average of Tenure is taken. *** and ** indicate significance at the 1%, and 5% levels, respectively.

Table 2 Team-managed and single-managed fund returns and demographic data

	Team-managed fund OARs			Single-managed fund OARs		
	(1)	(2)	(3)	(4)	(5)	(6)
Population	0.0535 [*] (0.080)			-0.0304 (0.488)		
Income		0.0161 (0.155)			-0.0329 (0.307)	
Education			0.0103 ^{**} (0.028)			-0.0031 (0.835)
Constant	Yes	Yes	Yes	Yes	Yes	Yes
R ² (%) Obs.	5.25 80	2.24 80	4.11 80	0.41 76	0.90 76	0.03 76

Panel A: Full sample of metro areas

Panel B: Metro areas with ten or more fund return observations

	Team-managed fund OARs			Single-managed fund OARs		
	(1)	(2)	(3)	(4)	(5)	(6)
Population	0.0328 ^{**} (0.029)			0.0375 (0.202)		
Income		0.0061 (0.448)			-0.0015 (0.864)	
Education			0.0051 (0.179)			0.0010 (0.860)
Constant	Yes	Yes	Yes	Yes	Yes	Yes
R ² (%) Obs.	4.75 60	0.68 60	2.43 60	3.20 60	0.02 60	0.05 60

This table shows the impact of managerial structure on the relation between fund performance and demographic variables. Fund performance is based on objective-adjusted returns (OARs) using the Morningstar Direct U.S. domestic equity mutual fund data from 1992 to 2010. All demographic data are from the 2000 U.S. Census. Population (log) is the population of U.S. metro areas. Income is the average per-capita income for the metro area. Education is the percent of population with university degree within the metro area. OAR (%, per month) is investment objective adjusted fund return, which is the difference between the average monthly net fund return for fund i in year t and the average monthly fund return of all funds in the matched investment objective in year t. Each regression also reports the p-values of coefficients, the number of observations, and the adjusted R^2 . Panel A uses fund return observations across all metro areas, while Panel B uses metro areas with at least ten fund return observations. All estimations are based on the Huber-White robust standard errors. ^{**} and ^{*} indicate significance at the 5%, and 10% levels, respectively.

Table 3 Performance differences in team-managed and single-managed funds across various metro areas

Panel A: Large pop	oulation areas			
Management	Statistic	α(4F)	α(5F)	CS
Team	Mean S.D.	0.0761 0.6039	0.0794 0.6552	0.2522 2.9519
Single	Mean S.D.	0.0544 0.6561	0.0574 0.7092	0.2558 3.2734
	Diff (T – S) p-value	0.0217 ^{**} (0.043)	0.0220 [*] (0.057)	-0.0036 (0.952)
Panel B: Small pop	ulation areas			
Management	Statistic	α(4F)	α(5F)	CS
Team	Mean S.D.	$0.0708 \\ 0.6077$	0.0787 0.6614	0.2169 2.9575
Single	Mean S.D.	0.0615 0.6685	0.0621 0.7143	0.4206 3.6817
	Diff (T – S) p-value	0.0093 (0.462)	0.0166 (0.227)	-0.2037 ^{**} (0.007)
Panel C: High educ	ation areas			
Management	Statistic	α(4F)	α(5F)	CS
Team	Mean	0.0757	0.0867	0.2865

	Diff (T – S) p-value	0.0309 ^{***} (0.003)	0.0359 ^{***} (0.002)	0.0178 (0.756)	
Single	Mean S.D.	0.0448 0.6592	$0.0508 \\ 0.7084$	0.2687 3.2271	
Team	Mean S.D.	0.0757 0.6040	0.0867 0.6535	0.2865 2.9906	

Panel C: Low education areas

Management	Statistic	α(4F)	α(5F)	CS
Team	Mean	0.0713	0.0691	0.1680
	S.D.	0.6075	0.6634	2.9010
Single	Mean	0.0795	0.0745	0.4362
	S.D.	0.6646	0.7165	3.8657
	Diff (T – S)	-0.0082	-0.0054	-0.2682 ^{***}
	p-value	(0.536)	(0.708)	(0.001)

Table 3 (continued)

This table shows the difference in fund performance between team-managed and single-managed funds metro areas of different size using the Morningstar Direct U.S. domestic equity mutual fund data from 1992 to 2010. Metro area population data are from the 2000 U.S. Census. Large population area funds have their advisors located within 50 miles of Boston, Chicago, Los Angeles, New York, Philadelphia, San Francisco, and Washington, D.C. α (4F), α (5F), are the Carhart and Pastor-Stambaugh alphas, respectively, while CS is the characteristic selectivity measure of Daniel, Grinblatt, Titman, and Wermers (1997). The p-values for the difference test between team-managed and single-managed funds, Diff (T –S), are in parentheses. ***, ** and * indicate significance at the 1%, 5%, and 10% levels, respectively.

Table 4Team management, metro area size, and fund performance

	L	Large population areas			Small population areas		
	α(4F)	α(5F)	CS	α(4F)	α(5F)	CS	
Team _{i,t-1}	0.0427 ^{**}	0.0558 ^{***}	0.1087 [*]	-0.0024	-0.0007	-0.0214	
	(0.016)	(0.000)	(0.059)	(0.877)	(0.967)	(0.802)	
Fund Size _{i,t-1}	-0.0223 ^{***}	-0.0195 ^{***}	-0.1046 ^{***}	-0.0271 ^{**}	-0.0201 [*]	-0.0389	
	(0.005)	(0.003)	(0.000)	(0.015)	(0.060)	(0.477)	
Fund Age _{i,t-1}	-0.0135	-0.0110	-0.1149 ^{**}	-0.0072	-0.0105	-0.1020	
	(0.249)	(0.417)	(0.016)	(0.498)	(0.327)	(0.286)	
Family Size _{i,t-1}	-0.0949 ^{***}	-0.0782 ^{***}	-0.2405 ^{***}	-0.0775 ^{***}	-0.0576	-0.3341 ^{***}	
	(0.000)	(0.000)	(0.009)	(0.003)	(0.007)	(0.003)	
Expenses _{i,t-1}	0.0481 ^{**}	0.0553 ^{**}	0.0266	0.0717^{**}	0.0619 [*]	0.3486 [*]	
	(0.024)	(0.024)	(0.764)	(0.044)	(0.060)	(0.084)	
Turnover _{i,t-1}	-0.0126	-0.0286	0.0452	-0.0031	-0.0110	0.0806	
	(0.482)	(0.161)	(0.554)	(0.884)	(0.634)	(0.281)	
Volatility _{i,t-1}	0.0042	-0.0155	-0.0046	-0.0578	-0.1055	-0.2187	
	(0.952)	(0.822)	(0.990)	(0.478)	(0.136)	(0.672)	
Flows _{i,t-1}	0.0026	0.0011	0.0281 ^{**}	-0.0064	-0.0031	-0.0178	
	(0.499)	(0.777)	(0.017)	(0.427)	(0.681)	(0.497)	
Tenure _{i,t-1}	-0.0135 [*]	-0.0042	0.0329	0.0033	0.0056	0.1348 ^{****}	
	(0.100)	(0.657)	(0.244)	(0.750)	(0.659)	(0.005)	
MBA _{i,t-1}	0.0571 ^{**}	0.0542 ^{**}	0.1181	-0.0166	0.0207	-0.2387	
	(0.036)	(0.016)	(0.230)	(0.647)	(0.627)	(0.222)	
Female _{i,t-1}	-0.0963 ^{***}	-0.0579 ^{**}	0.1003	-0.1036 ^{**}	-0.0768	-0.1810	
	(0.004)	(0.050)	(0.451)	(0.033)	(0.119)	(0.546)	
Constant	Yes	Yes	Yes	Yes	Yes	Yes	
Time \times Obj. FE Fund Family FE Metro Area FE $R^2(\%)$ Obs.	Yes Yes Yes 18.92 11,290	Yes Yes Yes 20.07 11,108	Yes Yes Yes 10.95 8,148	Yes Yes 20.25 8,323	Yes Yes 20.96 8,165	Yes Yes Yes 11.79 5,644	
Team (LPA-SPA) p-value	0.0498 ^{***} (0.010)	0.0538 ^{**} (0.015)	0.1623 [*] (0.067)				

Panel A: Full sample period

Table 4 (continued)

	1992-2004				2005-2010		
	α(4F)	α(5F)	CS	α(4F)	α(5F)	CS	
Team _{i,t-1} ×LPA _i	0.0569 ^{**}	0.0616 ^{**}	0.1247	0.0363 [*]	0.0471 ^{**}	0.0779 ^{***}	
	(0.026)	(0.014)	(0.247)	(0.077)	(0.031)	(0.000)	
$Team_{i,t-1} \times SPA_i$	0.0063	-0.0006	0.0549	-0.0135	0.0187	-0.0399	
	(0.776)	(0.982)	(0.695)	(0.615)	(0.512)	(0.740)	
Fund & Mgr. Controls	Yes	Yes	Yes	Yes	Yes	Yes	
Constant	Yes	Yes	Yes	Yes	Yes	Yes	
Time \times Obj. FE Fund Family FE Metro Area FE $R^2(\%)$ Obs.	Yes Yes Yes 24.38 9,297	Yes Yes 29.35 9,332	Yes Yes Yes 12.62 7,301	Yes Yes Yes 14.95 9,959	Yes Yes Yes 12.77 9,941	Yes Yes Yes 15.11 6,491	

Panel B: Sub-sample periods

This table shows the impact of management structure and fund location on fund performance using the Morningstar Direct U.S. domestic equity mutual fund data from 1992 to 2010. It reports the estimates from panel regressions of fund performance on Team and other controls. Metro area population data are from the 2000 U.S. Census. Large metro area funds have their advisors located within 50 miles of Boston, Chicago, Los Angeles, New York, Philadelphia, San Francisco, and Washington, D.C. Small metro areas are all other localities. The dependent variable includes three fund performance measures: $\alpha(4F)$, $\alpha(5F)$, and CS. $\alpha(4F)$, $\alpha(5F)$, are the Carhart and Pastor-Stambaugh alphas, respectively, while CS is the characteristic selectivity measure of Daniel, Grinblatt, Titman, and Wermers (1997). Independent variable of interest is Team dummy, which equals one if the fund has two (or more) fund managers and zero otherwise. Other independent variables are defined as in Table 1. Panel A shows the full sample period results. Team (LPA-SPA) gives the difference in the Team coefficient across metro areas and the corresponding p-value of the F-test for the equality of estimates. Panel B shows results for the 1992-2004 and 2005-2010 sub-periods. All regressions include time by investment objective, fund family, and metro area fixed effects, while the standard errors are clustered by fund and year. Also reported are the p-values of coefficients, the number of observations, and the adjusted R². ***, ** and * indicate significance at the 1%, 5%, and 10% levels, respectively.

Table 5Team management, metro area education level, and fund performance

	High education area			Low education area		
	α(4F)	α(5F)	CS	α(4F)	α(5F)	CS
Team _{i,t-1}	0.0419 ^{***}	0.0568 ^{***}	0.0865	-0.0115	-0.0190	-0.0911
	(0.007)	(0.002)	(0.129)	(0.564)	(0.269)	(0.248)
Fund & Mgr. Controls	Yes	Yes	Yes	Yes	Yes	Yes
Constant	Yes	Yes	Yes	Yes	Yes	Yes
Time \times Obj. FE	Yes	Yes	Yes	Yes	Yes	Yes
Fund Family FE	Yes	Yes	Yes	Yes	Yes	Yes
Metro Area FE	Yes	Yes	Yes	Yes	Yes	Yes
$R^2(\%)$	19.17	21.50	11.41	17.87	19.16	11.92
Obs.	11,665	11,677	8,739	7,593	7,598	5,055
Team (HEA-LEA) p-value	0.0567 ^{***} (0.002)	0.0718 ^{***} (0.004)	0.1561 (0.113)			

Panel A: Full sample period

Panel B: Sub-sample periods

	1992-2004				2005-2010		
	α(4F)	α(5F)	CS	α(4F)	α(5F)	CS	
Team _{i,t-1} ×HEA _i	0.0593 ^{**}	0.0610 ^{**}	0.1054	0.0325	0.0603^{*}	0.0526	
	(0.012)	(0.023)	(0.195)	(0.154)	(0.079)	(0.411)	
Team _{i,t-1} ×LEA _i	-0.0045	-0.0066	0.0820	-0.0083	-0.0025	-0.0061	
	(0.873)	(0.809)	(0.669)	(0.740)	(0.906)	(0.939)	
Fund & Mgr. Controls	Yes	Yes	Yes	Yes	Yes	Yes	
Constant	Yes	Yes	Yes	Yes	Yes	Yes	
Time \times Obj. FE	Yes	Yes	Yes	Yes	Yes	Yes	
Fund Family FE	Yes	Yes	Yes	Yes	Yes	Yes	
Metro Area FE	Yes	Yes	Yes	Yes	Yes	Yes	
$R^{2}(\%)$	24.39	29.36	12.62	14.98	12.80	15.12	
Obs.	9,298	9,333	7,302	9,960	9,942	6,492	

This table shows the impact of management structure and fund location on fund performance using the Morningstar Direct U.S. domestic equity mutual fund data from 1992 to 2010. It reports the estimates from panel regressions of fund performance on Team and other controls. Metro area population data are from the 2000 U.S. Census. Metro areas with high (low) education level, HEA (LEA), are those where percent of university degree holders is greater (lower) than 30.5%. The dependent variable includes four fund performance measures: $\alpha(4F)$, $\alpha(5F)$, and CS. $\alpha(4F)$, $\alpha(5F)$, are the Carhart and Pastor-Stambaugh alphas, respectively, while CS is the characteristic selectivity measure of Daniel, Grinblatt, Titman, and Wermers (1997). Independent variable of interest is Team dummy, which equals one if the fund has two (or more) fund managers and zero otherwise. Other independent variables are defined as in Table 1. Panel A shows the full sample period results, Panel B – for the 1992-2004 and 2005-2010 sub-periods. All regressions include time by investment objective, fund family, and metro area fixed effects, while the standard errors are clustered by fund and year. Also reported are the p-values of coefficients, the number of observations, and the adjusted R². ***, ** and * indicate significance at the 1%, 5%, and 10% levels, respectively.

Table 6Team, location, and fund performance across investment objectives

	Aggressive Growth and Growth			Growth & Income and Equity Income		
	α(4F)	α(5F)	CS	α(4F)	α(5F)	CS
Team _{i,t-1} ×LPA _i	0.0398 [*]	0.0534 ^{**}	0.1022	0.0477 ^{**}	0.0342	0.0663
	(0.062)	(0.013)	(0.129)	(0.066)	(0.214)	(0.604)
$Team_{i,t-1} \times SPA_i$	-0.0078	-0.0001	-0.0876	0.0341	0.0334	0.0904
	(0.735)	(0.997)	(0.269)	(0.351)	(0.397)	(0.560)
Fund & Mgr. Controls	Yes	Yes	Yes	Yes	Yes	Yes
Constant	Yes	Yes	Yes	Yes	Yes	Yes
Time \times Obj. FE Fund Family FE Metro Area FE $R^2(\%)$ Obs.	Yes Yes Yes 18.35 15,451	Yes Yes 20.48 15,471	Yes Yes Yes 10.58 11,331	Yes Yes 23.41 3,805	Yes Yes Yes 22.60 3,802	Yes Yes Yes 21.00 2,461
Team (LPA-SPA)	0.0476 ^{**}	0.0535 ^{**}	0.1898 ^{**}	0.0136	0.0008	-0.0241
p-value	(0.023)	(0.029)	(0.032)	(0.678)	(0.979)	(0.898)

This table shows the impact of management structure and fund location on fund performance depending on fund investment objective. The fund data is the Morningstar Direct U.S. domestic equity mutual fund data from 1992 to 2010. Metro area population data are from the 2000 U.S. Census. Large population metro area (LPA) funds have their advisors located within 50 miles of Boston, Chicago, Los Angeles, New York, Philadelphia, San Francisco, and Washington, D.C. Small population metro areas (SPA) are all other localities. The dependent variable includes four fund performance measures: $\alpha(4F)$, $\alpha(5F)$, and CS. $\alpha(4F)$, $\alpha(5F)$, are the Carhart and Pastor-Stambaugh alphas, respectively, while CS is the characteristic selectivity measure of Daniel, Grinblatt, Titman, and Wermers (1997). Independent variables of interest are two interactive terms, Team×LPA and Team×SPA, where Team is a dummy variable, which equals one if the fund has two (or more) fund managers and zero otherwise. Other independent variables are defined as in Table 1. All regression includes time by investment objective, fund family, and metro area fixed effects, while the standard errors are clustered by fund and year. Also reported are the p-values of coefficients, the number of observations, and the adjusted R². Team (LPA-SPA) gives the difference in the Team coefficient across metro areas and the corresponding p-value of the F-test for the equality of estimates. ***, *** and * indicate significance at the 1%, 5%, and 10% levels, respectively.

Table 7 Alternative performance evaluation measures

	Metro areas by population			Metro areas by education		
	α(3F)	α(4C)	RG	α(3F)	α(4C)	RG
Team _{i,t} ×LPA _i	0.0321 ^{**} (0.036)	0.0328 ^{**} (0.029)	0.0809 [*] (0.063)			
Team _{i,t} ×SPA _i	-0.0100 (0.653)	-0.0023 (0.882)	0.0979^{**} (0.043)			
Team _{i,t} ×HEA _i				0.0380 ^{**} (0.012)	0.0360 ^{**} (0.016)	0.1019 ^{***} (0.010)
Team _{i,t} ×LEA _i				-0.0227 (0.252)	-0.0101 (0.636)	0.0647 (0.206)
Fund & Mgr. Controls	Yes	Yes	Yes	Yes	Yes	Yes
Constant	Yes	Yes	Yes	Yes	Yes	Yes
Time × Obj. FE Fund Family FE Metro Area FE R ² (%) Obs.	Yes Yes Yes 18.00 19,264	Yes Yes Yes 18.13 19,253	Yes Yes Yes 20.82 19,639	Yes Yes Yes 18.01 19,266	Yes Yes Yes 18.15 19,255	Yes Yes 20.83 19,641

This table shows the impact of management structure and fund location on fund performance using the Morningstar Direct U.S. domestic equity mutual fund data from 1992 to 2010. It reports the estimates from panel regressions of fund performance on Team and other controls. Metro area population data are from the 2000 U.S. Census. Large population metro area (LPA) funds have their advisors located within 50 miles of Boston, Chicago, Los Angeles, New York, Philadelphia, San Francisco, and Washington, D.C. Small population metro areas (SPA) are all other localities. Metro areas with high (low) education level, HEA (LEA), are those where percent of university degree holders is greater (lower) than 30.5%. The dependent variable includes three fund performance measures: $\alpha(3F)$, $\alpha(4C)$, and RG. $\alpha(3F)$ is the Fama-French three-factor alpha; $\alpha(4C)$ is the conditional Carhart (1997) alpha, where the market beta is a function of the U.S. T-bill and term-spread; RG is the return gap measure of Kacperczyk, Sialm, and Zheng (2008). Independent variable of interest is Team dummy, which equals one if the fund has two (or more) fund managers and zero otherwise. Other independent variables are defined as in Table 1. All regressions include time by investment objective, fund family, and metro area fixed effects, while the standard errors are clustered by fund and year. Also reported are the p-values of coefficients, the number of observations, and the adjusted R². ***, *** and * indicate significance at the 1%, 5%, and 10% levels, respectively.

Table 8 Team formation/de-formation age, location, and fund performance

	Team formation type			Team de-formation type		
	α(4F)	α(5F)	CS	α(4F)	α(5F)	CS
Large metro areas						
New (1-3 years)	0.0369	0.0384	0.0645	0.0745	0.0664	0.1237
Old (4+ years)	0.0809	0.0825	0.3933	0.0362	0.0221	0.0189
Diff (Old-New) p-value	0.0440 ^{**} (0.044)	0.0439 [*] (0.067)	0.0328 ^{***} (0.006)	-0.0583 (0.264)	-0.0443 (0.235)	-0.1048 (0.492)
Small metro areas						
New (1-3 years ago)	0.0576	0.0499	0.3394	0.0410	0.0157	0.4187
Old (4+ years ago)	0.0609	0.0669	0.1782	0.0537	0.0440	0.2152
Diff (Old-New) p-value	0.0033 (0.894)	0.0170 (0.532)	-0.1612 (0.121)	0.0127 (0.769)	0.0283 (0.545)	-0.2035 (0.323)

Panel A: Same number of managers

Panel B: Same managers

	Team formation type			Team de-formation type		
	α(4F)	α(5F)	CS	α(4F)	α(5F)	CS
Large metro areas						
New (1-3 years)	0.0358	0.0310	0.0604	0.0846	0.0675	0.1216
Old (4+ years)	0.0896	0.0894	0.4291	0.0259	0.0148	0.0417
Diff (Old-New) p-value	0.0538 ^{**} (0.027)	0.0584 ^{**} (0.028)	0.0368 ^{***} (0.006)	-0.0587 [*] (0.087)	-0.0527 (0.161)	-0.0799 (0.604)
Small metro areas						
New (1-3 years)	0.0564	0.0521	0.3283	0.0481	0.0231	0.4798
Old (4+ years)	0.0603	0.0528	0.1951	0.0632	0.0546	0.2657
Diff (Old-New) p-value	0.0039 (0.889)	0.0007 (0.980)	-0.1332 (0.315)	0.0151 (0.729)	0.0315 (0.502)	-0.2140 (0.301)

This table shows the relation between the formation and de-formation of team-management and fund performance across metro areas of different size using the Morningstar Direct U.S. domestic equity mutual fund data from 1992 to 2010. Panel A shows the results for the same number of managers, while Panel B – for the same portfolio managers. Metro area population data are from the 2000 U.S. Census. Large metro area funds have their advisors located within 50 miles of Boston, Chicago, Los Angeles, New York, Philadelphia, San Francisco, and Washington, D.C. New (Old) team formation denotes funds that became team-managed up to three (four or more) years ago. Similarly, New (Old) team de-formation denotes funds that became single-managed up to three (four or more) years ago. $\alpha(4F)$, $\alpha(5F)$, are the Carhart and Pastor-Stambaugh alphas, respectively, while CS is the characteristic selectivity measure of Daniel, Grinblatt, Titman, and Wermers (1997). They are trimmed at the top and the bottom 1% of the distribution. The p-values for the difference test between old and new team formation/de-formation, Diff (Old-New), are in parentheses. ** and * indicate significance at the 5% and 10% levels, respectively.

Table 9Portfolio holdings across metro areas

		All years			
	(1)	(2)	(3)	1992-2004	2005-2010
$Team_{i,t-1} \times LPA_i$	0.0712 ^{***} (0.007)	0.0660 ^{**} (0.015)	0.0640 ^{**} (0.018)	0.0643 [*] (0.070)	0.0871 ^{***} (0.010)
$Team_{i,t-1} \times SPA_i$	0.1703^{***} (0.000)	0.1374 ^{***} (0.000)	0.1344 ^{****} (0.000)	0.0902 ^{**} (0.011)	0.1495 ^{***} (0.000)
Fund Controls Manager Controls Constant	Yes	Yes Yes	Yes Yes Yes	Yes Yes Yes	Yes Yes Yes
Time × Obj. FE Fund Family FE Metro Area FE $R^2(\%)$ Obs.	Yes Yes 40.25 25,285	Yes Yes Yes 49.27 18,684	Yes Yes Yes 49.31 18,652	Yes Yes 50.83 9,130	Yes Yes Yes 54.67 9,522
Team (LPA-SPA) p-value	-0.0991 ^{***} (0.005)	-0.0714 [*] (0.059)	-0.0704 [*] (0.064)	-0.0269 (0.609)	-0.0624 (0.143)

This table shows the effect of management structure on fund holdings using Morningstar Direct data on U.S. domestic equity mutual funds and Morningstar holdings data from 1992 to 2010. It shows the estimates from panel regressions of the logarithm of the percent of fund holdings on management structure and fund and manager controls. Large population metro area (LPA) funds have their advisors located within 50 miles of Boston, Chicago, Los Angeles, New York, Philadelphia, San Francisco, and Washington, D.C. Small population metro areas (SPA) are all other localities. The independent variable of interest is Team, which equals one if the fund has two (or more) fund managers and zero if the fund has only one fund manager at the end of calendar year. Other independent variables include fund and manager characteristics from Table 1. All regression specifications include time by investment objective, fund family, and metro area fixed effects, while the standard errors are clustered by fund and year. Also reported are the p-values of coefficients, the number of observations, and the adjusted R². Team (LPA-SPA) gives the difference in the Team coefficient across locations and the corresponding p-value of the F-test for the equality of estimates. ***, ** and * indicate significance at the 1%, 5%, and 10% levels, respectively.

Table 10					
Active shares and	managerial	structure	across	metro	areas

	All years				
	(1)	(2)	(3)	1992-2004	2005-2010
Team _{i,t-1} ×LPA _i	-0.0057 (0.502)	-0.0042 (0.606)	-0.0027 (0.599)	-0.0059 (0.545)	-0.0020 (0.869)
$Team_{i,t-1} \times SPA_i$	-0.0415 ^{***} (0.001)	-0.0382 ^{***} (0.002)	-0.0349 ^{***} (0.006)	-0.0350 [*] (0.064)	-0.0420 ^{***} (0.004)
Fund Controls Manager Controls Constant	Yes	Yes Yes	Yes Yes Yes	Yes Yes Yes	Yes Yes Yes
Time × Obj. FE Fund Family FE Metro Area FE R ² (%) Obs.	Yes Yes 40.34 20,900	Yes Yes Yes 44.31 16,491	Yes Yes Yes 44.59 16,455	Yes Yes Yes 45.49 8,603	Yes Yes Yes 52.63 7,852
Team (LPA-SPA) p-value	0.0358 ^{**} (0.012)	0.0340 ^{**} (0.016)	0.0322 ^{**} (0.022)	0.0291 (0.139)	0.0400 ^{**} (0.037)

This table shows the impact of management structure and fund location on active share. The fund data is the Morningstar Direct U.S. domestic equity mutual fund data from 1992 to 2010. The active share data is provided by Martijn Cremers. Metro area population data are from the 2000 U.S. Census. Large population metro area (LPA) funds have their advisors located within 50 miles of Boston, Chicago, Los Angeles, New York, Philadelphia, San Francisco, and Washington, D.C. Small population metro areas (SPA) are all other localities. Independent variables of interest are two interactive terms, Team×LPA and Team×SPA, where Team is a dummy variable, which equals one if the fund has two (or more) fund managers and zero otherwise. Other independent variables are defined as in Table 1. All regression includes time by investment objective, fund family, and metro area fixed effects, while the standard errors are clustered by fund and year. Also reported are the p-values of coefficients, the number of observations, and the adjusted R^2 . Team (LPA-SPA) gives the difference in the Team coefficient across locations and the corresponding p-value of the F-test for the equality of estimates. ***, ** and * indicate significance at the 1%, 5%, and 10% levels, respectively.



Figure 1. Average number of fund managers by metro area population. The figure shows the average number of fund managers for 65 U.S. metro areas by population (in millions). Each of the metro areas has both team-managed and single-managed funds. The sample includes all U.S. domestic equity mutual funds from 1992 to 2010 from the Morningstar Direct database. Team size is trimmed at four fund managers.



Figure 2. Relation between metro are population and education level. The figure shows the relation between the natural log of population and percent of university degree holders in 91 U.S. metro areas that have domestic equity mutual fund data in 1992-2010. The demographic data are from the 2000 U.S. Census. The mutual fund data are from Morningstar Direct.



Plot A: Percent of the same managers in a team after team formation



Plot B: Percent of the same single manager after team de-formation

Figure 3. Stability of managerial structure from the team formation or de-formation year. The figure shows the percent of the same portfolio managers in the managerial structure of fund each year after team formation (Plot A) and each year after team de-formation (Plot B). The sample includes all U.S. domestic equity mutual funds from 1992 to 2010 from the Morningstar Direct database.