Social Relationships and Mortality Risk: A Meta-Analytic Review

Timothy B. Smith
*Brigham Young University*, tbs@byu.edu

Julianne Holt-Lunstad

J. Bradley Layton

Follow this and additional works at: [https://scholarsarchive.byu.edu/facpub](https://scholarsarchive.byu.edu/facpub)

Part of the Counseling Psychology Commons, and the Special Education and Teaching Commons

Original Publication Citation
Holt-Lunstad, J., Smith, T. B., & Layton, B. (21). Social relationships and mortality risk: A meta-analytic review. PLoS Medicine, 7(7): e1316. doi:1.1371/journal.pmed.1316

BYU ScholarsArchive Citation
Smith, Timothy B.; Holt-Lunstad, Julianne; and Layton, J. Bradley, "Social Relationships and Mortality Risk: A Meta-Analytic Review" (2010). *Faculty Publications*. 94.
[https://scholarsarchive.byu.edu/facpub/94](https://scholarsarchive.byu.edu/facpub/94)

This Peer-Reviewed Article is brought to you for free and open access by BYU ScholarsArchive. It has been accepted for inclusion in Faculty Publications by an authorized administrator of BYU ScholarsArchive. For more information, please contact ellen_amatangelo@byu.edu.
Social Relationships and Mortality Risk: A Meta-analytic Review

Julianne Holt-Lunstad1*, Timothy B. Smith2*, J. Bradley Layton3

1 Department of Psychology, Brigham Young University, Provo, Utah, United States of America, 2 Department of Counseling Psychology, Brigham Young University, Provo, Utah, United States of America, 3 Department of Epidemiology, University of North Carolina at Chapel Hill, Chapel Hill, North Carolina, United States of America

Abstract

Background: The quality and quantity of individuals’ social relationships has been linked not only to mental health but also to both morbidity and mortality.

Objectives: This meta-analytic review was conducted to determine the extent to which social relationships influence risk for mortality, which aspects of social relationships are most highly predictive, and which factors may moderate the risk.

Data Extraction: Data were extracted on several participant characteristics, including cause of mortality, initial health status, and pre-existing health conditions, as well as on study characteristics, including length of follow-up and type of assessment of social relationships.

Results: Across 148 studies (308,849 participants), the random effects weighted average effect size was OR = 1.50 (95% CI 1.42 to 1.59), indicating a 50% increased likelihood of survival for participants with stronger social relationships. This finding remained consistent across age, sex, initial health status, cause of death, and follow-up period. Significant differences were found across the type of social measurement evaluated (p < 0.001); the association was strongest for complex measures of social integration (OR = 1.91; 95% CI 1.63 to 2.23) and lowest for binary indicators of residential status (living alone versus with others) (OR = 1.19; 95% CI 0.99 to 1.44).

Conclusions: The influence of social relationships on risk for mortality is comparable with well-established risk factors for mortality.

Please see later in the article for the Editors’ Summary.

Citation: Holt-Lunstad J, Smith TB, Layton JB (2010) Social Relationships and Mortality Risk: A Meta-analytic Review. PLoS Med 7(7): e1000316. doi:10.1371/journal.pmed.1000316

Academic Editor: Carol Brayne, University of Cambridge, United Kingdom

Received December 30, 2009; Accepted June 17, 2010; Published July 27, 2010

Copyright: © 2010 Holt-Lunstad et al. This is an open-access article distributed under the terms of the Creative Commons Attribution License, which permits unrestricted use, distribution, and reproduction in any medium, provided the original author and source are credited.

Funding: This research was generously supported by grants from the Department of Gerontology at Brigham Young University awarded to JHL and TBS and from TP Industrial, Inc awarded to TBS. The funders had no role in study design, data collection and analysis, decision to publish, or preparation of the manuscript.

Competing Interests: The authors have declared that no competing interests exist.

Abbreviations: CI, confidence interval; CVD, cardiovascular disease; OR, odds ratio

* E-mail: julianne_holt-lunstad@byu.edu

These authors contributed equally to this work.
Introduction

“Social relationships, or the relative lack thereof, constitute a major risk factor for health—rivaling the effect of well-established health risk factors such as cigarette smoking, blood pressure, blood lipids, obesity and physical activity.”

—House, Landis, and Umberson; Science 1988 [1]

Two decades ago a causal association between social relationships and mortality was proposed after a review of five large prospective studies concluded that social relationships predict mortality [1]. Following the publication of this provocative review, the number of prospective studies of mortality that included measures of social relationships increased exponentially. Although the inverse association between social relationships and nonsuicide mortality has received increased attention in research, neither major health organizations nor the general public recognize it as a risk factor for mortality. This may be due in part to the fact that the literature has become unwieldy, with wide variation in how social relationships are measured across a large number of studies and disappointing clinical trials [2]. “Social relationships” has perhaps become viewed as a fuzzy variable, lacking the level of precision and control that is preferred in biomedical research. Thus, the large corpus of relevant empirical research is in need of synthesis and refinement.

Current evidence also indicates that the quantity and/or quality of social relationships in industrialized societies are decreasing. For instance, trends reveal reduced intergenerational living, greater social mobility, delayed marriage, dual-career families, increased single-residence households, and increased age-related disabilities [3,4]. More specifically, over the last two decades there has been a three-fold increase in the number of Americans who report having no confidant—now the modal response [3]. Such findings suggest that despite increases in technology and globalization that would presumably foster social connections, people are becoming increasingly more socially isolated. Given these trends, understanding the nature and extent of the association between social relationships and mortality is of increased temporal importance.

There are two general theoretical models that propose processes through which social relationships may influence health: the stress buffering and main effects models [5]. The buffering hypothesis suggests that social relationships may provide resources (informational, emotional, or tangible) that promote adaptive behavioral or neuroendocrine responses to acute or chronic stressors (e.g., illness, life events, life transitions). The aid from social relationships thereby moderates or buffers the deleterious influence of stressors on health. From this perspective, the term social support is used to refer to the real or perceived availability of social resources [6]. The main effects model proposes that social relationships may be associated with protective health effects through more direct means, such as cognitive, emotional, behavioral, and biological influences that are not explicitly intended as help or support. For instance, social relationships may directly encourage or indirectly model healthy behaviors; thus, being part of a social network is typically associated with conformity to social norms relevant to health and self-care. In addition, being part of a social network gives individuals meaningful roles that provide self-esteem and purpose to life [7,8].

Social relationships have been defined and measured in diverse ways across studies. Despite striking differences, three major components of social relationships are consistently evaluated [5]: (a) the degree of integration in social networks [9], (b) the social interactions that are intended to be supportive (i.e., received social support), and (c) the beliefs and perceptions of support availability held by the individual (i.e., perceived social support). The first subconstruct represents the structural aspects of social relationships and the latter two represent the functional aspects. Notably, these different subconstructs are only moderately intercorrelated, typically ranging between $r = 0.20$ and $0.30$ [9,10]. While all three components have been shown to be associated with morbidity and mortality, it is thought that each may influence health in different ways [11,12]. Because it is presently unclear whether any single aspect of social relationships is more predictive than others, synthesis of data across studies using several types of measures of social relationships would allow for essential comparisons that have not been conducted on such a large scale.

Empirical data suggest the medical relevance of social relationships in improving patient care [13], increasing compliance with medical regimens [13], and promoting decreased length of hospitalization [14,15]. Likewise, social relationships have been linked to the development [16,17] and progression [18–21] of cardiovascular disease [22]—a leading cause of death globally. Therefore, synthesis of the current empirical evidence linking social relationships and mortality, along with clarifications of potential moderators, may be particularly relevant to public health and clinical practice for informing interventions and policies aimed at reducing risk for mortality.

To address these issues, we conducted a meta-analysis of the literature investigating the association between social relationships and mortality. Specifically, we addressed the following questions: What is the overall magnitude of the association between social relationships and mortality across research studies? Do structural versus functional aspects of social relationships differentially impact the risk for mortality? Is the association moderated by participant characteristics (age, gender, health status, cause of mortality) or by study characteristics (length of clinical follow-up, inclusion of statistical controls)? Is the influence of social relationships on mortality a gradient or threshold effect?

Methods

Identification of Studies

To identify published and unpublished studies of the association between social relationships and mortality, we used three techniques. First, we conducted searches of studies from January 1900 to January 2007 using several electronic databases: Dissertation Abstracts, HealthSTAR, Medline, Mental Health Abstracts, PsycINFO, Social Sciences Abstracts, Sociological Abstracts via SocioFile, Academic Search Premier, ERIC, and Family & Society Studies Worldwide. To capture the broadest possible sample of relevant articles, we used multiple search terms, including mortality, death, deceased, died, dead, and remained alive, which were crossed with search words related to social relationships, including the terms social and interpersonal linked to the following words: support, network, integration, participation, cohesion, relationships, capital, and isolation. To reduce inadvertent omissions, we searched databases yielding the most citations (Medline, PsycINFO) two additional times. Next, we manually examined the reference sections of past reviews and of studies meeting the inclusion criteria to locate articles not identified in the database searches. Finally, we sent solicitation letters to authors who had published three or more articles on the topic.

Inclusion Criteria

We included in the meta-analysis studies that provided quantitative data regarding individuals’ mortality as a function of social relationships, including both structural and functional
aspects [23]. Because we were interested in the impact of social relationships on disease, we excluded studies in which mortality was a result of suicide or injury. We also excluded studies in which the only measurement of social support was an intervention provided within the context of the study (e.g., support group), the source of social support was nonhuman (e.g., a pet or higher power), or the social support was provided to others (i.e., giving support to others or measures of others' benefit from the support provided) rather than to the individual tracked for mortality status.

We coded studies that included participant marital status as one of several indicators of social support, but we excluded studies in which marital status was the only indicator of social support. We also excluded studies in which the outcome was not explicitly and solely mortality (e.g., combined outcomes of morbidity/mortality). Reports with exclusively aggregated data (e.g., census-level statistics) were also excluded. Manuscripts coded were all written in English, which accounted for 98% of the total retrieved. See Figure 1 for additional details.

Data Abstraction

To increase the accuracy of coding and data entry, each article was initially coded by two raters. Subsequently, the same article was independently coded by two additional raters. Coders extracted several objectively verifiable characteristics of the studies: (a) the number of participants and their composition by age, gender, marital status, distress level, health status, and pre-existing health conditions (if any), as well as the percentage of smokers and percentage of physically active individuals, and, of course, the cause of mortality; (b) the length of follow up; (c) the research design; and (d) the aspect of social relationships evaluated.

Data within studies were often reported in terms of odds ratios (ORs), the likelihood of mortality across distinct levels of social relationships. Because OR values cannot be meaningfully aggregated, all effect sizes reported within studies were transformed to the natural log OR (lnOR) for analyses and then transformed back to OR for interpretation. When effect size data were reported in any metric other than OR or lnOR, we transformed those values using statistical software programs and macros (e.g., Comprehensive Meta-Analysis [24]). In some cases when direct statistical transformation proved impossible, we calculated the corresponding effect sizes from frequency data in matrices of mortality status by social relationship status. When frequency data were not reported, we recovered the cell probabilities from the reported ratio and marginal probabilities. When survival analyses (i.e., hazard ratios) were reported, we calculated the effect size from the associated level of statistical

---

Figure 1. Flow diagram.
doi:10.1371/journal.pmed.1000316.g001
Models were appropriate. We examined the remaining variance to confirm that random effects models do not 
[28]. In each analysis conducted, we models take such between-studies variation into account, whereas 
will estimate different population effect sizes. Random effects variables serve as moderators of the observed association between 
studies actually reviewed [27]. The assumptions made in this approach produces results that generalize beyond the sample of 
manuscript until consensus was obtained. A random effects coding pairs were resolved through further scrutiny of the 
agreement was quite high for categorical variables (mean Cohen's 
kappa = 0.73, SD = .14). Discrepancies across 
analyses, publication bias is unlikely to threaten the results. Based on these several 
results of both analyses failed to reach statistical significance 
and the alternative to that test recommended by Peters and 
unnecessary. Third, we calculated both Egger's regression test 
meaning that adjustment to the omnibus effect size was 
publication bias, but this analysis failed to reveal any studies that 
''trim and fill'' methodology described by Duval and Tweedie 
4,274, which is the theoretical number of unpublished studies with 
several analyses. First, we calculated the fail-safe N [177] to be 
4,274, which is the theoretical number of unpublished studies with effect sizes averaging zero (no effect) that would be needed to 
render negligible the omnibus results. Second, we employed the “trim and fill” methodology described by Duval and Tweedie 
[178,179] to estimate the number of studies missing due to 
publication bias, but this analysis failed to reveal any studies that would need to be created on the opposite side of the distribution, meaning that adjustment to the omnibus effect size was unnecessary. Third, we calculated both Egger's regression test and the alternative to that test recommended by Peters and colleagues [180] that is better suited to data in lnOR format. The results of both analyses failed to reach statistical significance (p>0.05). Finally, we plotted a contour-enhanced funnel plot (Figure 2) [181]. The data obtained from this meta-analysis were fairly symmetrical with respect to their own mean; fewer than ten studies were “missing” on the left side of the distribution that would have made the plot symmetrical. Based on these several analyses, publication bias is unlikely to threaten the results.

Moderation by Social Relationship Assessment, and by Participant and Study Characteristics

Given that structural versus functional components of social relationships may influence health in different ways [11,12], the high degree of heterogeneity observed in the omnibus results may have been due in part to differences between the components of social relationships evaluated within and across studies. Hence the remaining analyses separately evaluate effect sizes obtained from structural, functional, and combined (structural and functional) measures of social relationships. Table 2 provides definitions of the types and subtypes of social relationships evaluated.

Results

Statistically nonredundant effect sizes were extracted from 148 studies ([29–176]; see Table 1). Data were reported from 308,649 participants, with 51% from North America, 37% from Europe, 11% from Asia, and 1% from Australia. Across all studies, the average age of participants at initial evaluation was 63.9 years, and participants were evenly represented across sex (49% female, 51% male). Of the studies examined, 60% involved community samples, but 24% examined individuals receiving outpatient medical treatment, and 16% utilized patients in inpatient medical settings. Of studies involving patients with a pre-existing diagnosis, 44% were specific to cardiovascular disease (CVD), 96% to cancer, 9% to renal disease, and the remaining 11% had a variety of conditions including neurological disease. Research reports most often (81%) considered all-cause mortality, but some restricted evaluations to mortality associated with cancer (9%), CVD (8%), or other causes (2%). Participants were followed for an average of 7.5 years (SD = 7.1, range = 3 months to 38 years), with an average of 29% of the participants dying within each study's follow-up period.

Omnibus Analysis

Across 148 studies, the random effects weighted average effect size was OR = 1.50 (95% confidence interval [CI] = 1.42 to 1.59), which indicated a 50% increased likelihood of survival as a function of stronger social relations. Odds ratios ranged from 0.77 to 6.50, with substantial heterogeneity across studies (F = 81% [95% CI = 78% to 84%]; Q[147] = 790, p<0.001; t² = 0.07), suggesting that systematic effect size variability was unaccounted for. Thus factors associated with the studies themselves (e.g., publication status), participant characteristics (e.g., age, health status), and the type of evaluation of social relationships (e.g., structural social networks versus perceptions of functional social support) may have moderated the overall results. We therefore conducted additional analyses to determine the extent to which these variables moderated the overall results.

To assess the possibility of publication bias [177], we conducted several analyses. First, we calculated the fail-safe N [177] to be 4,274, which is the theoretical number of unpublished studies with effect sizes averaging zero (no effect) that would be needed to render negligible the omnibus results. Second, we employed the “trim and fill” methodology described by Duval and Tweedie [178,179] to estimate the number of studies missing due to publication bias, but this analysis failed to reveal any studies that would need to be created on the opposite side of the distribution, meaning that adjustment to the omnibus effect size was unnecessary. Third, we calculated both Egger’s regression test and the alternative to that test recommended by Peters and colleagues [180] that is better suited to data in lnOR format. The results of both analyses failed to reach statistical significance (p>0.05). Finally, we plotted a contour-enhanced funnel plot (Figure 2) [181]. The data obtained from this meta-analysis were fairly symmetrical with respect to their own mean; fewer than ten studies were “missing” on the left side of the distribution that would have made the plot symmetrical. Based on these several analyses, publication bias is unlikely to threaten the results.

Moderation by Social Relationship Assessment, and by Participant and Study Characteristics

Given that structural versus functional components of social relationships may influence health in different ways [11,12], the high degree of heterogeneity observed in the omnibus results may have been due in part to differences between the components of social relationships evaluated within and across studies. Hence the remaining analyses separately evaluate effect sizes obtained from structural, functional, and combined (structural and functional) measures of social relationships. Table 2 provides definitions of the types and subtypes of social relationships evaluated.

Social Relationships and Mortality

Significance, often derived from 95% confidence intervals (CIs). Across all studies we assigned OR values less than 1.00 to data indicative of increased mortality and OR values greater than 1.00 to data indicative of decreased mortality for individuals with relatively higher levels of social relationships.

When multiple effect sizes were reported within a study at the same point in time (e.g., across different measures of social relationships), we averaged the several values (weighted by standard error) to avoid violating the assumption of independent samples. In such cases, the aggregate standard error value for the lnOR were estimated on the basis of the total frequency data without adjustment for possible correlation among the averaged values. Although this method was imprecise, the manuscripts included in the meta-analysis did not report the information necessary to make the statistical adjustments, and we decided not to impute values given the wide range possible. In analyzing the data we used the shifting units of analysis approach [25] which minimizes the threat of nonindependence in the data while at the same time allowing more detailed follow-up analyses to be conducted (i.e., examination of effect size heterogeneity). When multiple reports contained data from the same participants (publications of the same database), we selected the report containing the whole sample and eliminated reports of subsamples. When multiple reports contained the same whole sample, we selected the one with the longest follow-up duration. When multiple reports with the same whole sample were of the same duration, we selected the one reporting the greatest number of measures of social relationships.

In cases where multiple effect sizes were reported across different levels of social relationships (i.e., high versus medium, medium versus low), we extracted the value with the greatest contrast (i.e., high versus low). When a study contained multiple effect sizes across time, we extracted the data from the longest follow-up period. If a study used statistical controls in calculating an effect size, we extracted the data from the model utilizing the fewest statistical controls so as to remain as consistent as possible across studies (and we recorded the type and number of covariates used within each study to run post hoc comparative analyses). We coded the research design used rather than estimate risk of individual study bias. The coding protocol is available from the authors.

The majority of information obtained from the studies was extracted verbatim from the reports. As a result, the inter-rater agreement was quite high for categorical variables (mean Cohen’s kappa = 0.73, SD = 0.13) and for continuous variables (mean intraclass correlation [26] = 0.80, SD = 14). Discrepancies across coding pairs were resolved through further scrutiny of the manuscript until consensus was obtained.

Aggregate effect sizes were calculated using random effects models following confirmation of heterogeneity. A random effects approach produces results that generalize beyond the sample of studies actually reviewed [27]. The assumptions made in this meta-analysis clearly warrant this method: The belief that certain variables serve as moderators of the observed association between social relationships and mortality implies that the studies reviewed will estimate different population effect sizes. Random effects models take such between-studies variation into account, whereas fixed effects models do not [28]. In each analysis conducted, we examined the remaining variance to confirm that random effects models were appropriate.
Table 1. Overview of the 148 studies included in the meta-analysis.

| Source | Total Number of Participants | Average Age at Intake | Location of Study | Study Length | Cause of Mortality | Social Relationship Measure | Original Statistic Metric | lnOR | Standard Error |
|--------|-------------------------------|-----------------------|-------------------|--------------|-------------------|-----------------------------|--------------------------|------|---------------|
| Ahern et al., 1990 [29] | 353 | 50 | USA | 1 y | All-cause | Functional | M & SD | 0.27 | 0.36 |
| Alter et al., 2006 [30] | 3,138 | 64 | Canada | 5 y | CVD | Combined | Chi | 0.06 | 0.15 |
| Anstey et al., 2002 [31] | 2,065 | 78 | Australia | 9 y | All-cause | Structural | Freq | 0.44 | 0.09 |
| Astrand et al., 1989 [32] | 391 | 50 | Sweden | 22 y | All-cause | Combined | OR | 0.00 | 0.18 |
| Avlund et al., 1998 [33] | 727 | 70 | Denmark | 11 y | All-cause | Combined | OR | 0.40 | 0.16 |
| Avlund et al., 2004 [34] | 565 | 75 | Denmark, Finland | 5 y | All-cause | Structural | OR | 0.54 | 0.22 |
| Barefoot et al., 2005 [35] | 3,109 | 58 | Denmark | 7 y | All-cause | Structural | p | 0.15 | 0.12 |
| Berkman and Syme, 1979 [36] | 4,765 | 47 | USA | 9 y | All-cause | Structural | Freq | 0.60 | 0.30 |
| Berkman et al., 2004 [37] | 3,495 | 45 | France | 10 y | All-cause | Structural | RR | 1.61 | 0.14 |
| Birket-Smith et al., 1989 [38] | 128 | 73 | Denmark | 1 y | All-cause | Structural | R | 0.37 | 0.33 |
| Blazer, 1982 [39] | 331 | 72 | USA | 2 y | All-cause | Combined | OR | 1.05 | 0.30 |
| Blazer et al., 2001 [40] | 3,664 | 73 | USA | 3 y | All-cause | Combined | OR | 0.15 | 0.10 |
| Bowling, 1989 [41] | 503 | 73 | UK | 6 y | All-cause | Structural | Chi | 0.51 | 0.16 |
| Brown et al., 2003 [42] | 846 | NR | USA | 5 y | All-cause | Combined | OR | 0.01 | 0.22 |
| Brummet et al., 2005 [43] | 2,711 | 62 | USA | 11 y | All-cause | Functional | p | 0.25 | 0.17 |
| Burg et al., 2005 [44] | 1,899 | 75 | USA | 2 y | All-cause | Combined | OR | 0.00 | 0.18 |
| Burns et al., 2005 [45] | 147 | 63 | Australia | 7 y | Cancer | Combined | Combin | 0.45 | 0.31 |
| Butow et al., 1999 [46] | 125 | 55 | Australia | 2 y | Cancer | Combined | p | 0.35 | 0.33 |
| Bygren et al., 1996 [47] | 12,675 | 43 | Sweden | 9 y | All-cause | Structural | Freq | 0.41 | 0.07 |
| Case et al., 1992 [48] | 1,195 | 59 | Canada, USA | 4 y | CVD | Structural | RR | 0.68 | 0.25 |
| Cassileth et al., 1988 [49] | 203 | 60 | USA | 8 y | Cancer | Structural | Combin | −0.03 | 0.26 |
| Ceria et al., 2001 [50] | 1,786 | 78 | USA | 6 y | All-cause | Structural | RR | 1.01 | 0.12 |
| Chacko et al., 1996 [51] | 94 | 53 | USA | 4 y | CVD | Functional | p | 0.92 | 0.39 |
| Christensen et al., 1999 [52] | 133 | 29 | USA | 58 y | All-cause | Combined | Chi | 0.98 | 0.32 |
| Christensen et al., 1994 [53] | 78 | 54 | USA | 5 y | All-cause | Functional | Chi | 0.98 | 0.44 |
| Cohen et al., 1987 [54] | 155 | 73 | USA | 3 y | All-cause | Structural | T | 0.65 | 0.30 |
| Colon et al., 1991 [55] | 100 | 30 | USA | 2 y | Cancer | Functional | Chi | 0.86 | 0.38 |
| Cormman et al., 2003 [56] | 4,049 | NR | Taiwan | 3 y | All-cause | Structural | OR | 0.17 | 0.06 |
| Coyne et al., 2001 [57] | 189 | 53 | USA | 4 y | CVD | Functional | RR | 0.99 | 0.26 |
| Cree et al., 2000 [58] | 558 | 82 | Canada | 4 m | All-cause | Functional | OR | 0.30 | 0.34 |
| Cuipers, 2001 [59] | 424 | 85 | Netherlands | 1 y | All-cause | Functional | OR | −0.10 | 0.31 |
| Dalgaard & Haheim, 1998 [60] | 1,002 | 46 | Norway | 17 y | All-cause | Structural | p | 0.23 | 0.15 |
| Devins et al., 1990 [61] | 97 | 40 | Canada | 4 y | Other | Structural | R | −0.025 | 0.38 |
| Dickens et al., 2004 [62] | 556 | 60 | UK | 1 y | CVD | Functional | p | 0.65 | 0.45 |
| Ell et al., 1992 [63] | 294 | 61 | USA | 6 y | 11m | All-cause | Combined | p | −0.15 | 0.21 |
| Eng et al., 2002 [64] | 16,242 | 55 | USA | 10 y | All-cause | Structural | RR | 0.42 | 0.06 |
| Engedal, 1996 [65] | 334 | 82 | Norway | 3 y | All-cause | Structural | M & SD | 0.62 | 0.20 |
| Farmer et al., 1986 [66] | 320 | 60 | USA | 4 y | 7m | All-cause | Combined | OR | 0.81 | 0.22 |
| Forster & Stoller, 1992 [67] | 363 | 74 | USA | 7 y | All-cause | Combined | LnOR | −0.20 | 0.22 |
| Frasure-Smith et al., 2000 [68] | 887 | 59 | Canada | 1 y | CVD | Functional | p | 0.09 | 0.12 |
| Frick et al., 2005 [69] | 99 | 55 | Germany | 3 y | 11m | Cancer | Combined | p | 0.21 | 0.35 |
| Fry and Debats, 2006 [70] | 380 | 75 | Canada | 5 y | 11m | All-cause | Combined | RR | 0.42 | 0.24 |
| Fuhrer et al., 1999 [71] | 3,777 | 76 | France | 5 y | All-cause | Combined | RR | 0.38 | 0.13 |
| Funch & Marshall, 1983 [72] | 208 | 51 | USA | 20 y | Cancer | Structural | Combin | 0.17 | 0.26 |
| Ganzini et al., 1997 [73] | 100 | 73 | USA | 2 y | 6m | All-cause | Combined | Combin | 0.15 | 0.25 |
| Gellert et al., 1993 [74] | 136 | 47 | USA | 10 y | Cancer | Functional | RR | −0.24 | 0.40 |
| Giles et al., 2005 [75] | 1,477 | 80 | Australia | 10 y | All-cause | Structural | p | 0.21 | 0.10 |
| Giraldi et al., 1997 [76] | 74 | 51 | Italy | 6 y | Cancer | Functional | M & SD | 0.14 | 0.43 |
Table 1. Cont.

| Source | Total Number of Participants | Average Age at Intake | Location of Study | Study Length | Cause of Mortality | Social Relationship Measure | Original Statistic Metric | lnOR | Standard Error |
|--------|-----------------------------|-----------------------|-------------------|--------------|-------------------|----------------------------|----------------------------|------|---------------|
| Glass et al., 1999 [77] | 1,380 | 72 | USA | 13 y | All-cause | Structural | RR | 0.42 | 0.20 |
| Goldman et al., 1995 [78] | 7,478 | 77 | USA | 6 y | All-cause | Structural | OR | 0.30 | 0.06 |
| Goodwin et al., 1996 [79] | 328 | 72 | USA | 10 y | All-cause | Structural | p | 0.62 | 0.20 |
| Gorkin et al., 1993 [80] | 1,146 | 61 | USA | 10 m | All-cause | Functional | Freq | 0.23 | 0.28 |
| Grand et al., 1990 [81] | 645 | 75 | France | 4 y | All-cause | Combined | OR | 0.40 | 0.22 |
| Greenfield et al., 2002 [82] | 5,092 | NR | USA | 11 y | All-cause | Structural | RR | 0.38 | 0.14 |
| Greenwood et al., 1995 [83] | 1,274 | 59 | UK | 4 y | All-cause | Structural | RR | 0.43 | 0.17 |
| Grodner et al., 1996 [84] | 110 | 63 | USA | 6 y | All-cause | Combined | M & SD | 0.50 | 0.35 |
| Gustafsson et al., 1998 [85] | 421 | 81 | Sweden | 6 y | All-cause | Structural | OR | 0.24 | 0.19 |
| Hall et al., 1993 [86] | 5,921 | 60 | Sweden | 11 y | CVD | Structural | OR | 0.23 | 0.15 |
| Helweg-Larsen, 2003 [87] | 6,617 | 44 | Denmark | 13 y | All-cause | Combined | RR | 0.74 | 0.05 |
| Herndon et al., 1999 [88] | 206 | 61 | USA | 4 y 2 m | Cancer | Functional | p | 0.16 | 0.26 |
| Hill et al., 2005 [89] | 3,050 | 78 | USA | 8 y | All-cause | Combined | p | 0.08 | 0.07 |
| Hirdes & Forbes, 1992 [90] | 259 | 45 | Canada | 20 y | All-cause | Combined | RR | 0.55 | 0.29 |
| Ho, 1991 [91] | 946 | 77 | China | 2 y | All-cause | Combined | RR | 0.55 | 0.24 |
| House et al., 1982 [92] | 2,754 | 52 | USA | 12 y | All-cause | Structural | Combin | 0.27 | 0.17 |
| Hummer et al., 1999 [93] | 21,204 | 43 | USA | 8 y | All-cause | Structural | Freq | 0.45 | 0.05 |
| Irbarren et al., 2005 [94] | 5,108 | 25 | USA | 16 y | All-cause | Structural | Combin | 0.60 | 0.21 |
| Irvine et al., 1999 [95] | 634 | 64 | Canada | 2 y | All-cause | Structural | RR | 0.01 | 0.32 |
| Iwasaki et al., 2002 [96] | 11,560 | 55 | Japan | 7 y | All-cause | Combined | RR | 0.22 | 0.11 |
| Johnson et al., 2005 [97] | 3,698 | 43 | USA | 5 y | All-cause | Combined | p | 0.18 | 0.10 |
| Johnson et al., 1996 [98] | 1,257 | 64 | Sweden | 14 y | CVD | Functional | RR | 0.21 | 0.15 |
| Jorm et al., 1991 [99] | 228 | 79 | Australia | 5 y | All-cause | Functional | M & SD | 0.24 | 0.24 |
| Juon et al., 2003 [100] | 1,091 | 6 | USA | 28 y | All-cause | Structural | OR | 0.60 | 0.35 |
| Jylhä and Aro, 1989 [101] | 936 | NR | Finland | 6 y 6 m | All-cause | Combined | p | 0.32 | 0.12 |
| Kaplan et al., 1988 [102] | 5,320 | 49 | Finland | 5 y | All-cause | Structural | OR | 0.75 | 0.18 |
| Kaplan et al., 1994 [103] | 2,501 | 53 | Finland | 5 y 11 m | All-cause | Combined | RR | 0.27 | 0.19 |
| Kawachi et al., 1996 [104] | 18,702 | 60 | USA | 4 y | All-cause | Structural | RR | 0.50 | 0.17 |
| Keller et al., 2003 [105] | 654 | 78 | USA | 10 y | All-cause | Structural | p | 0.53 | 0.14 |
| Kiey et al., 2000 [106] | 916 | 87 | USA | 4 y 6 m | All-cause | Structural | p | 0.23 | 0.12 |
| Kimmel et al., 2000 [107] | 174 | 54 | USA | 5 y | All-cause | Functional | p | 0.73 | 0.17 |
| Korten et al., 1999 [108] | 752 | 70 | Australia | 4 y | All-cause | Combined | Combin | 0.20 | 0.13 |
| Krause, 1997 [109] | 2,209 | 68 | UK | 11 y | All-cause | Combined | OR | –0.03 | 0.10 |
| Krause, 2006 [110] | 976 | 74 | USA | 3 y | All-cause | Combined | OR | 0 | 0.18 |
| Kronke et al., 2006 [111] | 2,835 | 59 | USA | 12 y | All-cause | Structural | RR | 0.45 | 0.22 |
| La Cour et al., 2005 [112] | 734 | 70 | Denmark | 20 y | All-cause | Structural | p | 0.45 | 0.14 |
| Lee & Rotheram-Borus, 2001 [113] | 307 | 38 | USA | 2 y 4 m | Other | Functional | p | 0.54 | 0.21 |
| Lehto et al., 2006 [114] | 101 | 54 | Finland | 9 y | Cancer | Functional | p | 0.97 | 0.38 |
| Lennartsson and Silverstein, 2001 [115] | 463 | 82 | Sweden | 4 y | All-cause | Structural | RR | 0.40 | 0.17 |
| Ljungquist et al., 1995 [116] | 956 | 70 | Sweden | 10 y | All-cause | Combined | OR | 1.03 | 0.16 |
| Lund et al., 2002 [117] | 1,265 | 60 | Denmark | 8 y | All-cause | Structural | p | 0.37 | 0.16 |
| Lund et al., 2000 [118] | 894 | 79 | Denmark | 8 y | All-cause | Structural | OR | 0.30 | 0.21 |
| Lyra and Heikkinen, 2006 [119] | 206 | 80 | Finland | 10 y | All-cause | Combined | Combin | 0.25 | 0.30 |
| Maier & Smith, 1999 [120] | 513 | 85 | Germany | 6 y | All-cause | Functional | Combin | 0.33 | 0.16 |
| Malmstrom et al., 2001 [121] | 22,236 | 47 | Sweden | 8 y | All-cause | Structural | RR | 0.30 | 0.07 |
| McClellan et al., 1993 [122] | 210 | 55 | USA | 1 y | All-cause | Functional | M & SD | 0.24 | 0.34 |
| Merlo et al., 2000 [123] | 491 | 68 | Sweden | 10 y | All-cause | Combined | Freq | 0.63 | 0.19 |
| Mertens et al., 1996 [124] | 1,869 | 62 | USA | 4 y | All-cause | Structural | M & SD | 0.56 | 0.08 |
| Source | Total Number of Participants | Average Age at Intake | Location of Study | Study Length | Cause of Mortality | Social Relationship Measure | Original Statistic Metric | lnOR | Standard Error |
|--------|-----------------------------|-----------------------|------------------|--------------|-------------------|---------------------------|------------------------|------|---------------|
| Morris et al., 1993 [125] | 91 | 60 | USA | 10 y | All-cause | Structural | T | 0.81 | 0.40 |
| Murata et al., 2005 [126] | 1,994 | 73 | Japan | 7 y 4 m | All-cause | Combined | p | 0.12 | 0.11 |
| Murgberg and Bru, 2001 [127] | 119 | 66 | Norway | 2 y | CVD | Combined | p | 0.27 | 0.34 |
| Musick et al., 2004 [128] | 3,617 | 47 | USA | 7 y 6 m | All-cause | Combined | R | 0.17 | 0.06 |
| Nakanishi and Tatare, 2000 [129] | 1,285 | 74 | Japan | 5 y 6 m | All-cause | Structural | | 0.26 | 0.10 |
| Nordentoft et al., 1993 [130] | 974 | 41 | Denmark | 10 y | All-cause | Structural | p | 0.42 | 0.12 |
| Olsen et al., 1991 [131] | 1,637 | 79 | Denmark | 15 y 6m | All-cause | Combined | p | 0.14 | 0.11 |
| Oman and Reed, 1998 [132] | 2,023 | 75 | USA | 5 y 7 m | All-cause | Structural | P | 0.20 | 0.11 |
| Oreill et al., 2000 [133] | 60 | 80 | UK | 3 y | All-cause | Combined | p | 0.62 | 0.48 |
| Orth-Gomer and Johnson, 1987 [134] | 17,433 | 49 | Sweden | 6 y | ? | Structural | RR | 1.31 | 0.07 |
| Orth-Gomer and Under, 1990 [135] | 147 | 57 | Sweden | 10 y | All-cause | Structural | T | 0.86 | 0.40 |
| Ostbye et al., 2006 [136] | 4,012 | 77 | USA | 10 y | All-cause | Combined | OR | 0.54 | 0.09 |
| Oxman et al., 1995 [137] | 232 | 76 | USA | 6 m | CVD | Combined | Combin | 0.33 | 0.46 |
| Parkerson and Gutman, 2000 [138] | 103 | 63 | USA | 1 y | All-cause | Structural | OR | 1.65 | 0.58 |
| Pennix et al., 1997 [139] | 2,829 | 70 | Netherlands | 3 y | All-cause | Combined | Freq | 0.30 | 0.15 |
| Rasulo et al., 2005 [140] | 1,734 | 81 | Denmark | 6 y | All-cause | Structural | p | 0.11 | 0.09 |
| Reuben et al., 1992 [141] | 259 | 73 | USA | 4 y 3 m | All-cause | Combined | R | 0.52 | 0.22 |
| Reynolds et al., 1994 [142] | 1,011 | 53 | USA | 5 y | Cancer | Combined | p | 0.19 | 0.17 |
| Rodriguez-Artalejo et al., 2006 [143] | 251 | 77 | Spain | 7 m | CVD | Structural | p | 0.17 | 0.33 |
| Rosengren et al., 1998 [144] | 717 | 50 | Sweden | 12 y | All-cause | Combined | Freq | 0.64 | 0.28 |
| Roy et al., 1996 [145] | 547 | 80 | USA | 4 y | All-cause | Structural | RR | 0.76 | 0.15 |
| Rozzini et al., 1991 [146] | 1,201 | 73 | Italy | 3 y | All-cause | Structural | Freq | 0.94 | 0.20 |
| Ruberman et al., 1984 [147] | 2,320 | 50 | USA | 3 y | All-cause | Structural | Chi | 0.39 | 0.08 |
| Rutledge et al., 2003 [148] | 7,524 | 71 | USA | 6 y | All-cause | Combined | RR | 0.53 | 0.05 |
| Rutledge et al., 2004 [149] | 503 | 59 | USA | 2 y 4 m | All-cause | Combined | M & SD | 0.99 | 0.37 |
| Saito-Nakaya et al., 2006 [150] | 238 | 62 | Japan | 7 y 6 m | All-cause | Combined | Freq | −0.07 | 0.35 |
| Schoenbach et al., 1986 [151] | 791 | 55 | Canada | 2 y | All-cause | Structural | Freq | 0.80 | 0.19 |
| Seeman et al., 1993 [152] | 1,420 | 74 | USA | 5 y | All-cause | Combined | p | 1.83 | 0.17 |
| Shahatamhasabei et al., 1992 [153] | 534 | 72 | UK | 8 y | All-cause | Combined | Chi | 0.40 | 0.16 |
| Shmotkin et al., 2003 [154] | 1,174 | 84 | Israel | 8 y | All-cause | Structural | p | −0.09 | 0.12 |
| Shye et al., 1995 [155] | 455 | 72 | USA | 15 y | All-cause | Structural | Freq | 0.80 | 0.21 |
| Silverstein and Bengston, 1991 [156] | 435 | 67 | USA | 14 y | All-cause | Combined | OR | 0.03 | 0.16 |
| Soler-Vila et al., 2003 [157] | 322 | 54 | USA | 10 y | All-cause | Combined | M & SD | 0.29 | 0.20 |
| Stavrakakis et al., 1998 [158] | 224 | 59 | Canada | 1 y | Cancer | Combined | Freq | 0.55 | 0.35 |
| Stek et al., 2005 [159] | 476 | 85 | Netherlands | 5 y | All-cause | Functional | p | 0.35 | 0.21 |
| Sturdy et al., 2002 [160] | 1,066 | 53 | UK | 5 y | All-cause | Structural | OR | 0.17 | 0.35 |
| Sugisawa et al., 1994 [161] | 1,943 | 69 | Japan | 3 y | All-cause | Combined | p | 0.03 | 0.19 |
| Sun and Liu, 2006 [162] | 7,938 | 92 | China | 2 y | All-cause | Structural | R | 0.67 | 0.04 |
| Temkin-Greener et al., 2004 [163] | 3,138 | 79 | USA | 2 y | All-cause | Combined | p | 0.21 | 0.10 |
| Thomas et al., 1997 [164] | 424 | 63 | Canada, USA | 3 y 11m | CVD | Functional | M & SD | 0.10 | 0.18 |
| Tucker et al., 1996 [165] | 1,077 | 12 | USA | 41 y | All-cause | Structural | p | 0.27 | 0.12 |
| Vaillant et al., 1998 [166] | 223 | 20 | USA | 25 y | All-cause | Combined | OR | 1.15 | 0.37 |
| Vogt et al., 1992 [167] | 2,396 | 47 | USA | 15 y | All-cause | Structural | p | 0.20 | 0.08 |
| Walter-Ginzburg et al., 2002 [168] | 1,340 | 83 | Israel | 8 y | All-cause | Combined | Freq | 0.23 | 0.11 |
| Waxler-Morrison et al., 1991 [169] | 118 | 45 | Canada | 4 y | Cancer | Structural | p | 0.27 | 0.36 |
| Wehls et al., 2005 [170] | 90 | 52 | USA | 9 y | Cancer | Structural | Combin | 0.61 | 0.40 |
**Structural aspects of social relationships.** Sixty-three studies had data exclusive to structural measures of social relationships (see Figure 3). Across these studies, the random effects weighted average effect size was OR = 1.57 (95% CI = 1.28 to 1.86), which value fell within the CI of the omnibus results reported previously. The heterogeneity across studies was still quite large ($I^2 = 84\%$ [95% CI = 80% to 87%]; $Q_{292} = 390$, $p<0.001$; $t = 0.07$), so we undertook metaregression with prespecified participant and study characteristics.

Metaregression is an analogue to multiple regression analysis for effect sizes. Its primary purpose is to ascertain which continuous and categorical (dummy coded) variables predict variation in effect size estimates. Using random effects weighted metaregression, we examined the simultaneous association with all variables entered into the model between effect sizes and prespecified participant and study characteristics (Table 3). To examine the most precise effect size estimates available and to increase the statistical power associated with this analysis, we shifted the unit of analysis [24] and extracted effect sizes within studies that were specific to the same variables and analytic procedures described previously. Sixty-three of these studies had data exclusive to structural measures of social relationships that is, if a study contained effect sizes from both structural and functional types of social relationships, we extracted the structural types for this analysis (with identical subtypes aggregated), which resulted in a total of 230 unique effect sizes across 116 studies. A total of 18% of the variance in these effect sizes was explained in the metaregression ($p<0.001$). As can be seen in Table 3, effect sizes based on data controlling for other variables were lower in magnitude than those based on raw data. Moreover, effect sizes differed in magnitude across the subtype of structural social relationships measured. Complex measures of social integration were associated with larger effect size values than measures of social participation. Binary measures of whether participants lived alone (yes/no) were associated with smaller effect size values. Average random effects weighted odds ratios for the various subtypes of social relationships are reported in Table 4.

**Functional aspects of social relationships.** Twenty-four studies had data exclusive to functional measures of social relationships (see Figure 4). Across these studies, the random effects weighted average effect size was OR = 1.46 (95% CI = 1.28 to 1.66), which value fell within the CI of the omnibus results reported previously. There was moderate heterogeneity across studies ($I^2 = 47\%$ [95% CI = 16% to 68%]; $Q_{25} = 44$, $p<0.01$; $t = 0.07$), so we conducted a random effects metaregression using the same variables and analytic procedures described previously. We extracted 87 unique effect sizes that were specific to measures of functional social relationships within 72 studies. A total of 16.5% of the variance in these effect sizes was explained in the metaregression, but the model did not reach statistical significance ($p = 0.46$). The results were not moderated by any of the specified participant characteristics (age, sex, initial health status, cause of mortality) or study characteristics (length of follow-up, geographic region, statistical controls).

**Combined assessments of social relationships.** Sixty-one studies had combined data of both structural and functional measures of social relationships (see Figure 5). Across these studies, the random effects weighted average effect size was OR = 1.44 (95% CI = 1.32 to 1.58). A large degree of heterogeneity characterized studies ($I^2 = 82\%$ [95% CI = 78% to 86%]; $Q_{29} = 337$, $p<0.001$; $t^2 = 0.09$), and we conducted a random effects metaregression using the same variables and analytic procedures described previously. We extracted 64 unique effect sizes that evaluated combined structural and functional measures of social relationships within 61 studies. The metaregression explained only 6.8% of the variance in these effect sizes, and the model failed to reach statistical significance ($p = 0.46$). The results were not moderated by any of the specified participant characteristics (age, sex, initial health status, cause of mortality) or study characteristics (length of follow-up, geographic region, statistical controls).

**Discussion**

Cumulative empirical evidence across 148 independent studies indicates that individuals’ experiences within social relationships...
ships and mortality may be generalized. Of death, suggesting that the association between social relations—
including age, sex, initial health status, follow-up period, and cause—
Results also remained consistent across a number of factors, 
quite large, rivaling that of well-established risk factors (Figure 6). 
survival. Thus, the magnitude of these findings may be considered 
yielded an even stronger association: a 91% increase in odds of 
relationships. Multidimensional assessments of social integration 
with a 50% increase in odds of survival as a function of social 
relationships (e.g., Table 2). For instance, while researchers may be 
tempted to use a simple single-item such as “living alone” as a 
proxy for social isolation, it is possible for one to live alone but 
have a large supportive social network and thus not adequately 
capture social isolation. We also found that social isolation had a 
similar influence on likelihood of mortality compared with other 
measures of social relationships. This evidence qualifies the notion 
of a threshold effect (lack of social relationships is the only 
detrimental condition); rather, the association appears robust 
across a variety of types of measures of social relationships.

This meta-analysis also provides evidence to support the 
directional influence of social relationships on mortality. Most of 
the studies (60%) involved community cohorts, most of whom 
would not be experiencing life-threatening conditions at the point 
of initial evaluation. Moreover, initial health status did not 
moderate the effect of social relationships on mortality. Although 
illness may result in poorer or more restricted social relationships 
(social isolation resulting from physical confinement), such that 
individuals closer to death may have decreased social support 
compared to healthy individuals, the findings from these studies 
dicate that general community samples with strong social 
relationships are likely to remain alive longer than similar 
individuals with poor social relations. However, causality is not 
easily established. One cannot randomly assign human partici-
pants to be socially isolated, married, or in a poor-quality 
relationship. A similar dilemma characterizes virtually all lifestyle 
risk factors for mortality: for instance, one cannot randomly assign

Table 2. Descriptive coding of the measures used to assess social relationships.

| Type of Measure          | Description                                                                 | Example of Measure                                          |
|--------------------------|-----------------------------------------------------------------------------|-------------------------------------------------------------|
| **Functional**           | Functions provided or perceived to be available by social relationships     |                                                             |
| Received support         | Self-reported receipt of emotional, informational, tangible, or belonging support | Inventory of Social Supportive Behaviors [213]               |
|                          |                                                                             | UCLA Social Support Interview [214,215]                     |
| Perceptions of social support | Perception of availability of emotional, informational, tangible, or belonging support if needed. | EPESE support questions [217]                               |
|                          |                                                                             | Malmo Social Support Scale [218]                            |
| Perception of loneliness | Feelings of isolation, disconnectedness, and not belonging                  | Loneliness Scale [221]                                     |
|                          |                                                                             | UCLA Loneliness Scale [222]                                |
| **Structural**           | The existence and interconnections among differing social ties and roles    |                                                             |
| Marital status           | married versus other                                                        | Binary item: Married yes, no                                |
|                          |                                                                             | Married, never married, divorced, separated, widowed        |
| Social networks          | network density or size, number of social contacts                          | Convoy measure [223]                                       |
|                          |                                                                             | Social Network List [224]                                  |
| Social integration       | Participation in a broad range of social relationships; including active engagement in a variety of social activities or relationships, and a sense of communality and identification with one’s social roles. | Malmo Influence, Contact, & Anchorage Measure [225]         |
|                          |                                                                             | Social Network Index [226,227]                              |
|                          |                                                                             | Social Participation Scale [92]                             |
| Complex measures of social integration | A single measure that assesses multiple components of social integration such as marital status, network size and network participation. | Social Network Index [36]                                   |
|                          |                                                                             | Social Network Questionnaire [228]                         |
|                          |                                                                             | Social Connections Index [102]                             |
|                          |                                                                             | Rand Social Health Battery [229]                           |
| Living alone             | Living alone versus living with others                                       | Binary item: yes, no                                        |
|                          |                                                                             | Number of people in household                              |
| Social isolation         | Pervasive lack of social contact or communication, participation in social activities, or confidant | Social Isolation Scale [82]                                |
| Multifaceted Measurement | Assessment of both structural and functional measures                      |                                                             |
|                          | Multiple measures obtained that assess more than one of the above conceptualizations. |                                                             |

doI:10.1371/journal.pmed.1000316.t002

significantly predict mortality. The overall effect size corresponds with a 50% increase in odds of survival as a function of social relationships. Multidimensional assessments of social integration yielded an even stronger association: a 91% increase in odds of survival. Thus, the magnitude of these findings may be considered quite large, rivaling that of well-established risk factors (Figure 6). Results also remained consistent across a number of factors, including age, sex, initial health status, follow-up period, and cause of death, suggesting that the association between social relationships and mortality may be generalized.

The magnitude of risk reduction varied depending on the type of measurement of social relationships (see Table 4). Social relationships were most highly predictive of reduced risk of mortality in studies that included multidimensional assessments of social integration. Because these studies included more than one type of social relationship measurement (e.g., network based inventories, marital status, etc.), such a measurement approach may better represent the multiple pathways (described earlier) by which social relationships influence health and mortality [182]. Conversely, binary evaluations of living alone (yes/no) were the least predictive of mortality status. The reliability and validity of measurement likely explains this finding, and researchers are encouraged to use psychometrically sound measures of social relationships (e.g., Table 2). For instance, while researchers may be tempted to use a simple single-item such as “living alone” as a
Figure 3. Forest plot of structural measures.
doi:10.1371/journal.pmed.1000316.g003
Table 3. Random effects metaregression for effect size estimates of structural social relationships.

| Variable                        | B     | SE    | p     | β     |
|---------------------------------|-------|-------|-------|-------|
| (Constant)                      | 0.535 | 0.238 | 0.02  | 0.00  |
| Participants’ average agea      | −0.002| 0.002 | 0.49  | −0.06 |
| Participant sex compositionb    |       |       |       |       |
| 100% Female                     | 0.038 | 0.066 | 0.57  | 0.04  |
| 100% Male                       | 0.049 | 0.068 | 0.48  | 0.05  |
| Participant initial healthc     | −0.103| 0.085 | 0.23  | −0.10 |
| Cause of mortalityd             |       |       |       |       |
| Cardiovascular disease          | 0.081 | 0.161 | 0.61  | 0.03  |
| Cancer                          | −0.208| 0.139 | 0.13  | −0.12 |
| Length of follow-up evaluation  | −0.003| 0.005 | 0.54  | −0.05 |

| Measure of social relationshipsa|       |       |       |       |
| Living alone                    | −0.265| 0.106 | 0.013 | −0.18 |
| Marital status                  | −0.097| 0.074 | 0.19  | −0.10 |
| Social isolation                | −0.144| 0.178 | 0.42  | −0.05 |
| Social networks                 | −0.050| 0.071 | 0.48  | −0.06 |
| Complex measures of integration | 0.255 | 0.095 | 0.007 | 0.20  |

| Geographic region of studyf     |       |       |       |       |
| Asia                            | 0.057 | 0.154 | 0.71  | 0.05  |
| Europe                          | 0.221 | 0.134 | 0.10  | 0.25  |
| North America                   | 0.057 | 0.134 | 0.69  | 0.07  |
| Statistically controlled estimateg| −0.147| 0.058 | 0.01  | −0.17 |

*Age at study initiation.
*Contrasted with reports in which males and females were combined.
*Individuals with a pre-existing medical condition contrasted with community samples.
*Contrasted with all cause and all other causes.
*Contrasted with measures of social participation; see Table 2 for descriptions of each kind of measure.
*Contrasted with all other world regions combined.
*Statistically controlled estimate based on raw data.
β, standardized beta; B, unstandardized beta; SE, standard error. doi:10.1371/journal.pmed.1000316.t003

Table 4. Weighted average effect sizes across different measures of social relationships.

| Type of Measure                      | k | OR     | 95% CI          |
|--------------------------------------|---|---------|-----------------|
| Functional                           |   |         |                 |
| Received social support              | 9 | 1.22    | [0.91, 1.63]    |
| Perceptions of social support        | 73| 1.35    | [1.22, 1.49]    |
| Loneliness (inversed)                | 8 | 1.45    | [1.08, 1.94]    |
| Structural                           |   |         |                 |
| Living alone (inversed)              | 17| 1.19    | [0.99, 1.44]    |
| Marital status (married versus other)| 62| 1.33    | [1.20, 1.48]    |
| Social isolation (inversed)          | 8 | 1.40    | [1.06, 1.86]    |
| Social networks                      | 71| 1.45    | [1.32, 1.59]    |
| Social integration                   | 45| 1.52    | [1.36, 1.69]    |
| Complex measures of social integration| 30| 1.91    | [1.63, 2.23]    |
| Combined structural and functional  | 67| 1.47    | [1.34, 1.60]    |

These analyses shifted the units of analysis, with distinct effect size estimates within studies used within different categories of measurement, such that many studies contributed more than one effect size but not more than one per category of measurement.
OR, odds ratio, transformed from random effects weighted lnOR.
doi:10.1371/journal.pmed.1000316.t004

individuals to be smokers or nonsmokers. Despite such challenges, “smoking represents the most extensively documented cause of disease ever investigated in the history of biomedical research” [183]. The link between social relationships and mortality is currently much less understood than other risk factors; nonetheless there is substantial experimental, cross-sectional, and prospective evidence linking social relationships with multiple pathways associated with mortality (see [182] for review). Existing models for reducing risk of mortality may be substantially strengthened by including social relationship factors.

Notably, the overall effect for social relationships on mortality reported here may be a conservative estimate. Many studies included in the meta-analysis utilized single item measures of social relations, yet the magnitude of the association was greatest among those studies utilizing complex assessments. Moreover, because many studies statistically adjusted for standard risk factors, the effect may be underestimated, since some of the impact of social relationships on mortality may be mediated through such factors (e.g., behavior, diet, exercise). Additionally, most measures of social relations did not take into account the quality of the social relationships, thereby assuming that all relationships are positive. However, research suggests this is not the case, with negative social relationships linked to greater risk of mortality [184,185]. For instance, marital status is widely used as a measure of social integration; however, a growing literature documents its divergent effects based on level of marital quality [186,187]. Thus the effect of positive social relationships on risk of mortality may actually be much larger than reported in this meta-analysis, given the failure to account for negative or detrimental social relationships within the measures utilized across studies.

Other possible limitations of this review should be acknowledged. Statistical controls (e.g., age, sex, physical condition, etc.) employed by many of the studies rule out a number of potentially confounding variables that might account for the association between social relationships and mortality. However, studies used an inconsistent variety of controlling variables, and some reports involved raw data (Table 1). Although effect size magnitude was diminished by the inclusion of statistical controls only within the data obtained by measures of structural social relationships (but not functional or combined measures), future research can better specify which variables are most likely to impact the overall association. It must also be acknowledged that existing data...
primarily represent research conducted in North America and Western Europe. Although we found no differences across world region, future reviews inclusive of research written in all languages (not only English) with participants better representing other world regions may yield better estimates across populations.

Approximately two decades after the review by House and colleagues [1], a generation of empirical research validates their initial premise: Social relationships exert an independent influence on risk for mortality comparable with well-established risk factors for mortality (Figure 6). Although limited by the state of current investigations and possible omission of pertinent reports, this meta-analysis provides empirical evidence (nearly 30 times the number of studies previously reported) to support the criteria for considering insufficient social relationships a risk factor of mortality (i.e., strength and consistency of association across a wide range of studies, temporal ordering, and gradient of response) [188]. The magnitude of the association between social relationships and mortality has now been established, and this meta-analysis provides much-needed clarification regarding the social relationship factor(s) most predictive of mortality. Future research can shift to more nuanced questions aimed at (a) understanding the causal pathways by which social participation promotes health, (b) refining conceptual models, and (c) developing effective intervention and prevention models that explicitly account for social relations.

Some steps have already been taken identifying the psychological, behavioral, and physiological pathways linking social relationships to health [5,182,189]. Social relationships are linked to better health practices and to psychological processes, such as stress and depression, that influence health outcomes in their own right [190]; however, the influence of social relationships on health cannot be completely explained by these processes, as social relationships exert an independent effect. Reviews of such findings suggest that there are multiple biologic pathways involved (physiologic regulatory mechanisms, themselves intertwined) that in turn influence a number of disease endpoints [182,191–193]. For instance, a number of studies indicate that social support is linked to better immune functioning [194–197] and to immune-mediated inflammatory processes [198]. Thus interdisciplinary work and perspective will be important in future studies given the complexity of the phenomenon.

Perhaps the most important challenge posed by these findings is how to effectively utilize social relationships to reduce mortality risk. Preliminary investigations have demonstrated some risk

![Figure 4. Forest plot of functional measures.](doi:10.1371/journal.pmed.1000316.g004)
Figure 5. Forest plot of combined measures. 
doi:10.1371/journal.pmed.1000316.g005
reduction through formalized social interventions [199]. While the evidence is mixed [2,6], it should be noted that most social support interventions evaluated in the literature thus far are based on support provided from strangers; in contrast, evidence provided in this meta-analysis is based almost entirely on naturally occurring social relationships. Moreover, our analyses suggest that received support is less predictive of mortality than social integration (Table 4). Therefore, facilitating patient use of naturally occurring social relations and community-based interventions may be more successful than providing social support through hired personnel, except in cases in which patient social relations appear to be detrimental or absent. Multifaceted community-based interventions may have a number of advantages because such interventions are socially grounded and include a broad cross-section of the public. Public policy initiatives need not be limited to those deemed “high risk” or those who have already developed a health condition but could potentially include low- and moderate-risk individuals earlier in the risk trajectory [200]. Overall, given the significant increase in rate of survival (not to mention quality of life factors), the results of this meta-analysis are sufficiently compelling to promote further research aimed at designing and evaluating interventions that explicitly account for social relationship factors across levels of health care (prevention, evaluation, treatment compliance, rehabilitation, etc.).

**Conclusion**

Data across 308,849 individuals, followed for an average of 7.5 years, indicate that individuals with adequate social relationships have a 50% greater likelihood of survival compared to those with poor or insufficient social relationships. The magnitude of this effect is comparable with quitting smoking and it exceeds many well-known risk factors for mortality (e.g., obesity, physical inactivity). These findings also reveal significant variability in the predictive utility of social relationship variables, with multidimensional assessments of social integration being optimal when assessing an individual’s risk for mortality and evidence that social isolation has a similar influence on mortality to other measures of social relationships. The overall effect remained consistent across a number of factors, including age, sex, initial health status, follow-up period, and cause of death, suggesting that the association between social relationships and mortality may be general, and efforts to reduce risk should not be isolated to subgroups such as the elderly.

To draw a parallel, many decades ago high mortality rates were observed among infants in custodial care (i.e., orphanages), even when controlling for pre-existing health conditions and medical treatment [201–204]. Lack of human contact predicted mortality. The medical profession was stunned to learn that infants would die without social interaction. This single finding, so simplistic in hindsight, was responsible for changes in practice and policy that markedly decreased mortality rates in custodial care settings. Contemporary medicine could similarly benefit from acknowledging the data: Social relationships influence the health outcomes of adults.

Physicians, health professionals, educators, and the public media take risk factors such as smoking, diet, and exercise seriously; the data presented here make a compelling case for social relationship factors to be added to that list. With such recognition, medical evaluations and screenings could routinely include variables of social well-being; medical care could
recommend if not outright promote enhanced social connections; hospitals and clinics could involve patient support networks in implementing and monitoring treatment regimens and compliance, etc. Health care policies and public health initiatives could likewise benefit from explicitly accounting for social factors in efforts aimed at reducing mortality risk. Individuals do not exist in isolation; social factors influence individuals’ health through cognitive, affective, and behavioral pathways. Efforts to reduce mortality via social relationship factors will require innovation, yet innovation already characterizes many medical interventions that extend life at the expense of quality of life. Social relationship-based interventions represent a major opportunity to enhance not only the quality of life but also survival.

Supporting Information

**Alternative Language Abstract S1** Abstract translated into Japanese by Hideko Cannell.
Found at: doi:10.1371/journal.pmed.1000316.s001 (0.02 MB DOC)

**Alternative Language Abstract S2** Abstract translated into Spanish by Rod Vea.
Found at: doi:10.1371/journal.pmed.1000316.s002 (0.03 MB DOC)

**References**

1. House JS, Landis KR, Umberson D (1988) Social relationships and health. Science 241: 540–545.
2. Berkman LF, Blumenthal J, Bur M, et al. (2003) Effects of treating depression and low perceived social support on clinical events after myocardial infarction: the Enhancing Recovery in Coronary Heart Disease Patients (ENRICHD) Randomized Trial. JAMA 289: 3106–3116.
3. McPherson M, Smith-Lovin L (2006) Social isolation in America: Changes in core discussion networks over two decades. Am Sociol Rev 71: 353–375.
4. Putnam RD (2000) Bowling Alone: The collapse and revival of American community. New York, NY, US: Simon & Schuster.
5. Cohen S, Gottlieb BH, Underwood LG (2000) Social Relationships and Health. In: Cohen S, Underwood LG, Gottlieb BH, eds. Measuring and intervening in social support. New York: Oxford University Press. pp 3–25.
6. Cohen S, Gottlieb BH, Underwood LG (2001) Social relationships and health: challenges for measurement and intervention. Adv Mind Body Med 17: 129–141.
7. Cohen S (2004) Social relationships and health. Am Psychol 59: 676–684.
8. Thoits PA (1983) Multiple identities and psychological well-being: A reformulation and test of the social isolation hypothesis. Am Sociol Rev 48: 174–187.
9. Brissett I, Cohen S, Seeman TE (2000) Measuring social integration and social networks. In: Cohen S, Underwood LG, Gottlieb BH, eds. Social support measurement and intervention. A guide for health and social scientists. New York, NY, US: Oxford University Press. pp 83–65.
10. Reinhardt JP, Boerner K, Horowitz A (2006) Good to have but not to use: Differential impact of perceived and received support on well-being. J Soc Pers Relat 23: 117–129.
11. Lakey B, Cohen S (2000) Social support theory and measurement. In: Cohen S, Underwood LG, Gottlieb BH, eds. Social support measurement and intervention: A guide for health and social scientists. New York, NY, US: Oxford University Press. pp 29–52.
12. Cohen S, Gottlieb BH, Underwood LG (2000) Social relationships and health. In: Cohen S, Underwood LG, Gottlieb BH, eds. Social support measurement and intervention: A guide for health and social scientists. New York, NY, US: Oxford University Press. pp 3–25.
13. DiMatteo MR (2004) Social support and patient adherence to medical treatment: a meta-analysis. Health Psychol 23: 207–218 (2004).
14. Murphy BM, Elliott PC, Le Grande MR, Higgins RO, Ernest CS, et al. (2000) Living alone predicts 30-day hospital readmission after coronary artery bypass graft surgery. Eur J Cardiovasc Prev Rehabil 15: 210–215.
15. Lett HS, Blumenfeld JA, Babayak MA, Catellier DJ, Carney RM, et al. (2007) Social support and prognosis in patients at increased psychosocial risk recovering from myocardial infarction. Health Psychol 26: 418–427.
16. Knox SS, Adelman A, Ellison RC, Amstek DK, Siegman K, et al. (2000) Hostility, social support, and carotid artery atherosclerosis in the National Heart, Lung, and Blood Institute Family Heart Study. Am J Cardiol 86: 1086–1089.
17. Kop WJ, Berman DS, Granser H, Wong ND, Miranda-Peats R, et al. (2005) Social network and coronary artery calcification in asymptomatic individuals. Psychosom Med 67: 543–552.
18. Brummett BH, Barefoot JC, Siegler IC, Clapp-Channing NE, Lytle BL, et al. (2001) Characteristics of socially isolated patients with coronary artery disease who are at elevated risk for mortality. Psychosom Med 63: 267–272.
19. Wang HX, Mintlemann MA, Lezinewer C, Orth-Gomer K (2006) Depressive symptoms, social isolation, and progression of coronary artery atherosclerosis: the Stockholm Female Coronary Angiography Study. Psychother Psychosom 75: 96–102.
20. Wang HX, Mintlemann MA, Orth-Gomer K (2005) Influence of social support on progression of coronary artery disease in women. Soc Sci Med 60: 599–607.
21. Angerer P, Siebert U, Kothny W, Muhlbauer D, Mutha H, et al. (2000) Impact of social support, cynical hostility and anger expression on progression of coronary atherosclerosis. J Am Coll Cardiol 36: 1751–1758.
22. Knox SS, Uvnas-Moberg K (1998) Social isolation and cardiovascular disease: an atherosclerotic pathway? Psychoneuroendocrinology 23: 877–890.
23. Cohen S, Wills TA (1996) Stress, social support, and the buffering hypothesis. Psychol Bull 98: 510–537.
24. Borenstein M, Hedges L, Rothstein H (2005) Comprehensive Meta-Analysis Version 2, Biostat, Englewood NJ.
25. Cooper H (1988) Synthesizing research: A guide for literature reviews 3rd ed. Thousand Oaks: Sage.
26. Shrouf PE, Elris J (1979) Intra-class correlations: Uses in assessing rater reliability. Psychol Bull 86: 420–428.
27. Hedges LV, Vevea J (1998) Fixed- and random-effects models in meta-analysis. Psychological Methods 3: 486–504.
28. Mosteller F, Colditz GA (1996) Understanding research synthesis (meta-analysis). Annual Review of Public Health 17: 1–23.
29. Ahern D, Geokin L, Anderson J, Tietnay G, Halteman A, et al. (1990) Bio behavioral variables and mortality or cardiac arrest in the cardiac arrhythmia pilot study (CAPS). Am J Cardiol 66: 59–62.
30. Alter DA, Chong A, Austin PC, Mustard C, Iron K, et al. (2006) Socioeconomic status and mortality after acute myocardial infarction. Ann Intern Med 144: 82–93.
31. Anstey JK, Laszcz MA (2002) Mortality risk varies according to gender and change in depressive status in very old adults. Psychosom Med 64: 75: 96–102.
32. Astrand NE, Hanson BS, Jacobsen SO (1989) Job demands, job decision latitude, job support, and social network factors as predictors of mortality in a Swedish pulp and paper company. Br J Ind Med 46: 334–340.
33. Avlund K, Dasumgaard MT, Holstein BE (1998) Social relations and mortality. An eleven year follow-up study of 70 year-old men and women in Denmark. Soc Sci Med 47: 635–643.
34. Avlund K, Land R, Holstein BE, Due P, Sakari-Rantala R (2004) The impact of structural and functional characteristics of social relations as determinants of functional decline. J Gerontol 59B: 444–451.
35. Baresfoot JC, Grobaek M, Jenson G, Schohr, Prescot E (2005) Social network diversity and risks of ischemic heart disease and total mortality: Findings from the Copenhagen City Heart Study. Ann J Epidemiol 161: 960–967.

**Text S1** PRISMA checklist.
Found at: doi:10.1371/journal.pmed.1000316.s003 (0.06 MB DOC)

**Text S2** Review protocol.
Found at: doi:10.1371/journal.pmed.1000316.s004 (0.05 MB DOC)

**Acknowledgments**

We would like to thank Jennie Bingham, Wendy Birmingham, Anne Brevon, Hoku Conkllk, Shavna Rae Cope, Katie Dyon, Stacie Fraire, Jeffrey Gale, Karen Gouchnour, Angela Salas Hamaker, Adan Howard, Brian Mead, Esther Rawlings, Kelly Smith, Effie Thacker, and Hiroko Umeda for their assistance with coding. We would also like to thank Bert Uchino, University of Utah, and Teresa Seeman, UCLA, for their helpful feedback on an earlier version of this paper.

**Author Contributions**

ICMJE criteria for authorship read and met: JHL TS JBL. Agree with the manuscript’s results and conclusions: JHL TS JBL. Designed the experiments/the study: TS. Analyzed the data: JHL TS JBL. Collected data/did experiments for the study: JHL TS JBL. Wrote the first draft of the paper: JHL TS. Contributed to the writing of the paper: JHL TS JBL.
Cohen CI, Teresi J, Holmes D (1987) Social networks and mortality in an
American men: The Honolulu Heart Program. J Am Geriatr Soc 49: 725–731.

Corman JC, Goldman N, Gleis DA, Weinstein M, Chang M (1993) Social ties
and perceived support: Two dimensions of social relationships and health
among the elderly in Taiwan. J Aging Health 15: 616–644.

Coyne JC, Rohrbough MJ, Shaham V, Sonneja J, Nicholas JM, et al. (2001)
Prognostic importance of marital quality for survival of congestive heart failure.
Am J Cardiol Ill: 526–529.

Cree M, Soskline CL, Bechse E, Hornig J, McElhaney JE (2000) Mortality and
mortality in elderly persons: a case for multiple, independent pathways. J
Gerontol: Medical Sciences, 56A: M505–M509.

Crawford BR, Walsh WP, Lusk EJ (1988) Psychosocial correlates of cancer
survival from patients undergoing autologous peripheral blood stem cell transplantation. J Psychosom Res 36: 531–540.

Czechowski DM (1990) Social support as a predictor of mortality of coronary patients: Effects
of smoking, sedentary behavior, and depressive symptoms. Psychosom Med 67: 40–45.

Czechowski DM (1990) Social support as a predictor of mortality of coronary patients: Effects
of smoking, sedentary behavior, and depressive symptoms. Psychosom Med 67: 40–45.

Czechowski DM (1990) Social support as a predictor of mortality of coronary patients: Effects
of smoking, sedentary behavior, and depressive symptoms. Psychosom Med 67: 40–45.

Czechowski DM (1990) Social support as a predictor of mortality of coronary patients: Effects
of smoking, sedentary behavior, and depressive symptoms. Psychosom Med 67: 40–45.

Czechowski DM (1990) Social support as a predictor of mortality of coronary patients: Effects
of smoking, sedentary behavior, and depressive symptoms. Psychosom Med 67: 40–45.

Czechowski DM (1990) Social support as a predictor of mortality of coronary patients: Effects
of smoking, sedentary behavior, and depressive symptoms. Psychosom Med 67: 40–45.

Czechowski DM (1990) Social support as a predictor of mortality of coronary patients: Effects
of smoking, sedentary behavior, and depressive symptoms. Psychosom Med 67: 40–45.

Czechowski DM (1990) Social support as a predictor of mortality of coronary patients: Effects
of smoking, sedentary behavior, and depressive symptoms. Psychosom Med 67: 40–45.

Czechowski DM (1990) Social support as a predictor of mortality of coronary patients: Effects
of smoking, sedentary behavior, and depressive symptoms. Psychosom Med 67: 40–45.

Czechowski DM (1990) Social support as a predictor of mortality of coronary patients: Effects
of smoking, sedentary behavior, and depressive symptoms. Psychosom Med 67: 40–45.

Czechowski DM (1990) Social support as a predictor of mortality of coronary patients: Effects
of smoking, sedentary behavior, and depressive symptoms. Psychosom Med 67: 40–45.

Czechowski DM (1990) Social support as a predictor of mortality of coronary patients: Effects
of smoking, sedentary behavior, and depressive symptoms. Psychosom Med 67: 40–45.

Czechowski DM (1990) Social support as a predictor of mortality of coronary patients: Effects
of smoking, sedentary behavior, and depressive symptoms. Psychosom Med 67: 40–45.

Czechowski DM (1990) Social support as a predictor of mortality of coronary patients: Effects
of smoking, sedentary behavior, and depressive symptoms. Psychosom Med 67: 40–45.

Czechowski DM (1990) Social support as a predictor of mortality of coronary patients: Effects
of smoking, sedentary behavior, and depressive symptoms. Psychosom Med 67: 40–45.

Czechowski DM (1990) Social support as a predictor of mortality of coronary patients: Effects
of smoking, sedentary behavior, and depressive symptoms. Psychosom Med 67: 40–45.

Czechowski DM (1990) Social support as a predictor of mortality of coronary patients: Effects
of smoking, sedentary behavior, and depressive symptoms. Psychosom Med 67: 40–45.
Jorm AF, Henderson AS, Kay DWK, Jacomb PA (1991) Mortality in relation to dementia, depression, and social integration in an elderly community sample. Int J Geriatr Psychiatry 6: 5–11.

Joss H, Esminger ME, Veerman M (1989) Childhood adversity and later mortality in an urban African American cohort. Am J Public Health 89: 2044–2046.

Julf M, Aro S (1989) Social ties and survival among the elderly in Tampere, Finland. Int J Epidemiol 18: 158–173.

Kiemeney LA, Salonen JT, Cohen RD, Brand RJ, Syme SL, et al. (1989) Social connections and mortality from all causes and from cardiovascular disease: Prospective evidence from eastern Finland. Am J Epidemiol 28: 370–380.

Kaplan GA, Wilson TW, Cohen RD, Kauhanen J, Wu M, et al. (1994) Social functioning and overall mortality: Prospective evidence from the Kuopio ischemic heart disease risk factor study. Epidemiology 5(6): 495–500.

Kawachi I, Colditz GA, Ascherio A, Rimm EB, Giovannucci E (1996) A prospective study of social networks in relation to total mortality and cardiovascular disease in men in the USA. J Epidemiol Community Health 50: 245–251.

Keller BK, Magnuson TM, Cermin PA, Stoner JA, Potter JF (2003) The significance of social network in a geriatric assessment population. Aging Clin Exp Res 15: 312–317.

Kleider DK, Simon SE, Jones RN, Morris JN (2000) The protective effect of social engagement on mortality in long-term care. J Am Geriatr Soc 48: 1367–1372.

Könnönen PE, Peterson RA, Wehrs KL, Shidler N, Simmons SJ, et al. (2000) Dyadic relationship conflict, gender, and mortality in Urban hemodialysis patients. J Am Soc Nephrol 11(8): 1518–1525.

Korten AE, Jorm AF, Jiao Z, Leteete L, Jacomb PA, et al. (1999). Health, cognitive, and psychosocial factors as predictors of mortality in an elderly community sample. J Epidemiol Community Health 53: 83–88.

Krause N (1997) Received support, anticipated support, social class, and mortality. Res Aging 19: 307–422.

Kroenke CH, Kubzansky LD, Schernhammer ES, Holmes MD, Kawachi I (2004) Social networks, social support, and longevity: A 14-year follow-up study among elderly in the elderly. Psychosom Med 66: 5–15.

Kroenke CH, Kubzansky LD, Schernhammer ES, Holmes MD, Kawachi I (2000) Social networks, social support, and survival after breast cancer diagnosis. J Clin Oncol 24: 1105–1111.

La Cour P, Ahlund K, Schulz Greger K (2005) Religion and survival in a secular region. A twenty year follow-up of 734 Danish adults born in 1914. Soc Sci Med 62: 137–164.

Lee M, Rotheram-Borus MJ (2003) Challenges associated with increased religious involvement among young adults with heart failure. J Am Soc Nephrol 14: 1261–1267.

Lehto US, Ojanen M, Dyba T, Aromaa A, Kellokumpu-Lehtinen P (2006) Psychological predictors of survival in localized breast cancer. J Clin Oncol 24: 1105–1111.

Lehner US, Ojanen M, Dyba T, Aromaa A, Kellokumpu-Lehtinen P (2007) Baseline psychosocial predictors of survival in localized breast cancer. Br J Cancer 94: 1245–1252.

Lemonsart C, Sewerstien M (2003) Does engagement with life enhance survival of elderly people in Sweden? The role of social and leisure activities. J Gerontol 58B: S335–S342.

Ljungquist B, Berg S, Steen B (1995) Prediction of survival in 70-year olds. Acta Oncol 34: 294–297.

Lund R, Due P, Modvig J, Holstein BE, Damsgaard MT, et al. (2000) Religion, religiosity and mortality in a national sample. J Health Soc Behav 41: 198–213.

Lund R, Due P, Modvig J, Holstein BE (2003) The association of depression with 10-year post stroke mortality. Am J Psychiatry 160: 124–129.

Mager D, Laakso M, Pasanen M, Reunanen A, Tuomilehto J (1994) Social networks, social support, and myocardial infarction risk in Finland. J Chronic Dis 47: 487–493.

Majer D, Samul RJ (1999) Psychological predictors of mortality in old age. J Gerontol: Series B. Psychol Sci & Soc Sciences 54B: 44–51.

Malmstrom M, Johannson S, Sundquist J (2001) A hierarchical analysis of long-term illness and mortality in socially deprived areas. Soc Sci Med 53: 263–275.

McClain W, Staniszewski DJ, Anson CA (1993) Social support and subsequent mortality among patients with end-stage renal disease. J Am Soc Nephrol 4: 1028–1034.

Merlo J, Ostergren P, Mannson N, Hanson BS, Ranstam J (2000) Mortality in elderly men with low psychosocial coping resources using anxiolytic-hypnotic drugs. 1403–4948: 26: 294–297.

Mertens JR, Moos RH, Brennan PL (1996) Alcohol consumption, life context, and coping predict mortality among late-middle-aged drinkers and former drinkers. Alcohol Clin Exp Res 20: 310–319.

Morris PLP, Robinson RG, Andrzejewski P, Samuels J, Price TR (1993) Association of depression with 10-year post stroke mortality. Am J Psychiatry 150: 124–129.
154. Shmotkin D, Blumstein T, Modan B (2003) Beyond keeping active: Concomitants of being a volunteer in old-old age. Psychol Aging 18: 602–607.
155. Shye D, Mullooly JP, Freeborn DK, Pepe CR (1995) Gender differences in the relationship between social network support and mortality: A longitudinal study of an elderly cohort. Soc Sci Med 41: 935–947.
156. Silverstein M, Bengtson VL (1991) Do close parent-child relations reduce the mortality risk of older parents? J Health Soc Behav 32: 382–395.
157. Soler-Vilà H, Karl SV, Jones BA (2003) Prognostic significance of psychosocial factors in African-American and White breast cancer patients: A population based study. Cancer 98: 1299–1308.
158. Stavisky KM, Donner AP, Kincade JE, Stewart MA (1988) The effect of psychosocial factors on lung cancer mortality at one year. J Clin Epidemiol 41: 1159–1168.
159. Stuckle M, Vinkers DJ, Giussele J, Beckman ATT, Van der Mast RC (2005) Is depression in old age fatal only when people feel lonely? The Am J Psychiatry 162: 178–180.
160. Sturdy PM, Victor CR, Anderson HR, Bland JM, Butland BK (2002) Psychological, social and health behavior risk factors for deaths certified as asthma: A national case-control study. Thorax 57: 1034–1039.
161. Sugawa H, Liang J, Liu X (1994) Social networks, social support, and mortality among older people in Japan. J Gerontol 49: S3–S13.
162. Sun R, Liu Y (2006) Mortality of the oldest old in China: The role of social and solitary customary activities. J Aging Health 18: 37–55.
163. Temkin-Greener H, Bajorska A, Peterson DR, Kunitz SJ, Gross D, et al. (2004) Social support and risk-adjusted mortality in a frail, older population. Med Care 42: 779–780.
164. Thomas SA, Friedmann E, Wimbush F, Schron E (1997) Psychosocial factors and survival in the Cardiac Arrhythmia Suppression Trial (CAST): A reexamination. Am J Crit Care 6: 116–126.
165. Tucker JS, Friedman HS, Wingard DL, Schwartz JE (1996). Marital status at midlife as a predictor of longevity: alternative explanations to the protective effects of marriage. Health Psychol 15: 94–101.
166. Vaillant GE, Meyer SE, Mukaamal K, Soldz S (1998) Are social supports in midlife a cause or a result of successful physical ageing? Psychol Med 28: 1159–1168.
167. Vogt TM, Mullooly JP, Ernst E, Pope CR, Hollis JF (1992) Social networks as predictors of ischemic heart disease, cancer, stroke, and hypertension: incidence, survival, and mortality. J Clin Epidemiol 45: 639–666.
168. Walter-Ginzburg A, Blumstein T, Chetrit A, Modan B (2002) Social factors and mortality in old-old in Israel: The CALAS study. J Gerontol 57: S308–S318.
169. Weisheit R, Simmerman SJ, Mizrahi E, Zhment MT, Hsu ME, et al. (2005) Dependent social relationships predict overall survival in stages II and III breast carcinoma patients. J Psychosom Res 59: 299–306.
170. Weing L, Lappas G, Wilhelmsen L (2000) Independent importance of psychosocial factors for progress after myocardial infarction. J Intern Med 249: 629–639.
171. Weism L, Larson B, Svardudd K, Tibblin B, Tibblin G (1992) Social network and activities in relation to mortality from cardiovascular diseases, cancer, and other causes: A 17-year follow up of the Study of Men Born in 1913 and 1923. Journal of Epidemiology and Community Health 46: 127–132.
172. Wilkins K (2003) Social support and mortality in seniors. Health Rep 14: 21–34.
173. Woloshin S, Schwartz LM, Tosteson ANA, Chang CH, Wright B, et al. (1997) Perceived adequacy of tangible social support and health outcomes in patients with coronary artery disease. J Gen Intern Med 12: 613–618.
174. Yasuda N, Zimmerman SL, Hawkes W, Fredman L, Hebel JR, et al. (1997) Relation of social network characteristics to 3-year mortality among young-old versus old-old White women in an urban community. Am J Epidemiol 145: 516–523.
175. Zuckerman DM, Kasl SV, Ostfeld AM (1984) Psychosocial predictors of mortality among middle-aged women. Am J Epidemiol 119: 526–536.
176. Zuckerman DM, Kasl SV, Ostfeld AM (1984) Psychosocial predictors of mortality among middle-aged women. Am J Epidemiol 119: 526–536.
177. Dunkel-Schetter C, Feinstein L, Call J (1987) UCLA Social Support Inventory. Unpublished manuscript, University of California, Los Angeles.
178. Dunkel-Schetter C, Folkman S, Lazarus RS (1987) Social support and mortality: the Framingham Offspring Study. Psychosom Med 69: 309–513.
179. Dunkel-Schetter C, Greenland S, Lash TL (2000) Modern Epidemiology. Philadelphia: Lippincott Williams & Wilkins.
180. Uchino BN, Cacioppo JT, Kiecolt-Glaser JK (1996) The relationship between social support and physiological processes: A review with emphasis on underlying mechanisms and implications for health. Psychol Bull 119: 498–531.
181. Roszanski A, Blumenthal JA, Kaplan J (1999) Impact of psychological factors on the pathogenesis of cardiovascular disease and implications for therapy. Circulation 99: 2192–2211.
182. Uchino BN, Hult-Lanstad J, Uno D, Campo R, Reblin M (2007) The Social Neurosciences of Relationships: An Examination of Health-Relevant Pathways. Social neuroscience: Integrating biological and psychological explanations of social behavior. New York, NY, US: Guilford Press. pp 474–492.
183. Uchino BN, Uno D, Hult-Lanstad J (1999) Social support, physiological processes, and health. J Behav Med 22: 1–14.
184. Roszanski A, Blumenthal JA, Kaplan J (1999) Impact of psychological factors on the pathogenesis of cardiovascular disease and implications for therapy. Circulation 99: 2192–2211.
217. Seeman TE, Berkman LF (1988) Structural characteristics of social support networks and their relationship with social support in the elderly: Who provides support. Soc Sci Med 26: 737–794.
218. Hanson BS, Ostergren P-O, Elmstahl S, Isacsson S-O, Ranstam J (1997) Reliability and validity assessments of measures of social networks, social support and control — results from the Malmo Shoulder and Neck Study. Scand J Soc Med 25: 249–257.
219. Sarason IG, Levine HM, Basham RB, Sarason BR (1983) Assessing social support: The social support questionnaire. J Pers Soc Psychol 44: 217–139.
220. Cohen S, Hoberman HM (1983) Positive events and social support as buffers of life change stress. J Appl Soc Psychol 13: 99–125.
221. De Jong-Gierveld J, Kamphuis F (1985) The development of a Rasch-type loneliness scale. Appl Psych Meas 9: 299–299.
222. Russell D, Peplau LA, Cutrona CE (1980) The revised UCLA Loneliness Scale: Concurrent and discriminant validity evidence. J Pers Soc Psychol 39: 472–480.
223. Kahn RL, Antonucci TC (1980) Convoys over the life course: Attachment, roles and social support. In Life span development and behavior Baltes PB, Brim 0, eds. New York, NY: Academic Press. pp 253–286.
224. Hirsch BJ (1979) Psychological dimensions of social networks: A multi-method analysis. Am J Commun Psychol. pp 163–277.
225. Hanson BS, Ostergren P-O, Elmstahl S, Isacsson S-O, Ranstam J (1997) Reliability and validity assessments of measures of social networks, social support and control — results from the Malmo Shoulder and Neck Study. Scand J Soc Med 25: 249–257.
226. Cohen S (1991) Social supports and physical health. In: Greene AL, Cummings M, Karraker KH, eds. Life-Span Developmental Psychology: Perspectives on Stress and Coping. Hillsdale, NJ: Erlbaum Associates.
227. Cohen S, Doyle WJ, Skoner DP, Rabin BS, Gwaltney JM (1997) Social ties and susceptibility to the common cold. JAMA 277: 1940–1944.
228. Kahn RL, Antonucci TC (1984) Social supports of the elderly: Family, friends, professionals (Refort No. AGO 01632). Bethesda, MD: National Institute on Aging.
229. Donald CA, Ware JE (1984) The measurement of social support. Res Commun Mental Health 4: 334–335.

Social Relationships and Mortality
Editors’ Summary

Background. Humans are naturally social. Yet, the modern way of life in industrialized countries is greatly reducing the quantity and quality of social relationships. Many people in these countries no longer live in extended families or even near each other. Instead, they often live on the other side of the country or even across the world from their relatives. Many also delay getting married and having children. Likewise, more and more people of all ages in developed countries are living alone, and loneliness is becoming increasingly common. In the UK, according to a recent survey by the Mental Health Foundation, 10% of people often feel lonely, a third have a close friend or relative who they think is very lonely, and half think that people are getting lonelier in general. Similarly, across the Atlantic, over the past two decades there has been a three-fold increase in the number of Americans who say they have no close confidants. There is reason to believe that people are becoming more socially isolated.

Why Was This Study Done? Some experts think that social isolation is bad for human health. They point to a 1988 review of five prospective studies (investigations in which the characteristics of a population are determined and then the population is followed to see whether any of these characteristics are associated with specific outcomes) that showed that people with fewer social relationships die earlier on average than those with more social relationships. But, even though many prospective studies of mortality (death) have included measures of social relationships since that first review, the idea that a lack of social relationships is a risk factor for death is still not widely recognized by health organizations and the public. In this study, therefore, the researchers undertake a systematic review and meta-analysis of the relevant literature to determine the extent to which social relationships influence mortality risk and which aspects of social relationships are most predictive of mortality. A systematic review uses predefined criteria to identify all the research on a given topic; a meta-analysis uses statistical methods to combine the results of several studies.

What Did the Researchers Do and Find? The researchers identified 148 prospective studies that provided data on individuals’ mortality as a function of social relationships and extracted an “effect size” from each study. An effect size quantifies the size of a difference between two groups—here, the difference in the likelihood of death between groups that differ in terms of their social relationships. The researchers then used a statistical method called “random effects modeling” to calculate the average effect size of the studies expressed as an odds ratio (OR)—the ratio of the chances of an event happening in one group to the chances of the same event happening in the second group. They report that the average OR was 1.5. That is, people with stronger social relationships had a 50% increased likelihood of survival than those with weaker social relationships. Put another way, an OR of 1.5 means that by the time half of a hypothetical sample of 100 people has died, there will be five more people alive with stronger social relationships than people with weaker social relationships. Importantly, the researchers also report that social relationships were more predictive of the risk of death in studies that considered complex measurements of social integration than in studies that considered simple evaluations such as marital status.

What Do These Findings Mean? These findings indicate that the influence of social relationships on the risk of death are comparable with well-established risk factors for mortality such as smoking and alcohol consumption and exceed the influence of other risk factors such as physical inactivity and obesity. Furthermore, the overall effect of social relationships on mortality reported in this meta-analysis might be an underestimate, because many of the studies used simple single-item measures of social isolation rather than a complex measurement. Although further research is needed to determine exactly how social relationships can be used to reduce mortality risk, physicians, health professionals, educators, and the media should now acknowledge that social relationships influence the health outcomes of adults and should take social relationships as seriously as other risk factors that affect mortality, the researchers conclude.

Additional Information. Please access these Web sites via the online version of this summary at http://dx.doi.org/10.1371/journal.pmed.1000316.

- The Mental Health America Live Your Life Well page includes information about how social relationships improve both mental and physical health
- The Mental Health Foundation, a UK charity, has information on loneliness and mental health; its report “The Lonely Society?” can be downloaded from this page
- The Mayo Clinic has information on social support as a way to manage stress
- The Pew Research Foundation has information on technology and social isolation
- Wikipedia has a page on social isolation (note that Wikipedia is a free online encyclopedia that anyone can edit; available in several languages)