Self-reported health without clinically measurable benefits among adult users of multivitamin and multimineral supplements: a cross-sectional study

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INTRODUCTION

Consumption of multivitamins (MVs) and multiminerals (MMs) (together MVMs) as dietary supplements is widespread in the general US adult population, with some reports estimating 33% of Americans regularly take MVMs.1-3 While MVM supplementation is warranted for some individuals at high risk because of disease-related deficiency,5 the consumption of non-prescription, over-the-counter MVMs has not produced robust evidence for the wide-ranging health benefits expected by the general adult population. Likewise, large randomised clinical trials that evaluate MVMs at different doses, across both men and women at various ages, have failed to demonstrate benefit in prevention of chronic diseases. The Physicians’ Health Study II (PHS II), a randomised placebo-controlled clinical trial of low-dose daily MVM use in older male physicians, found no reduction in major cardiovascular disease

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ABSTRACT

Objective Multiple clinical trials fail to identify clinically measurable health benefits of daily multivitamin and multimineral (MVM) consumption in the general adult population. Understanding the determinants of widespread use of MVMs may guide efforts to better educate the public about effective nutritional practices. The objective of this study was to compare self-reported and clinically measurable health outcomes among MVM users and non-users in a large, nationally representative adult civilian non-institutionalised population in the USA surveyed on the use of complementary health practices.

Design Cross-sectional analysis of the effect of MVM consumption on self-reported overall health and clinically measurable health outcomes.

Participants Adult MVM users and non-users from the 2012 National Health Interview Survey (n=21 603).

Primary and secondary outcome measures Five psychological, physical, and functional health outcomes: (1) self-rated health status, (2) needing help with routine needs, (3) history of 10 chronic diseases, (4) presence of 19 health conditions in the past 12 months, and (5) Kessler 6-item (K6) Psychological Distress Scale to measure non-specific psychological distress in the past month.

Results Among 4933 adult MVM users and 16 670 adult non-users, MVM users self-reported 30% better overall health than non-users (adjusted OR 1.31; 95% CI 1.17 to 1.46; false discovery rate adjusted p<0.001). There were no differences between MVM users and non-users in history of 10 chronic diseases, number of present health conditions, severity of current psychological distress on the K6 Scale and rates of needing help with daily activities. No effect modification was observed after stratification by sex, education, and race.

Conclusions MVM users self-reported better overall health despite no apparent differences in clinically measurable health outcomes. These results suggest that widespread use of multivitamins in adults may be a result of individuals’ positive expectation that multivitamin use leads to better health outcomes or a self-selection bias in which MVM users intrinsically harbour more positive views regarding their health.

Strengths and limitations of the study

► This study links better self-reported health, absence of clinically measurable benefits, and multivitamin and multimineral supplement use in the same population.

► Data are derived from a large, national survey across the USA.

► Results have broad implications for public health and the multibillion dollar supplement industry.

► Cross-sectional study design precludes the demonstration of a causal relationship between self-reported health and multivitamin and multimineral supplements.

► Self-reported health can be inherently biased and confounding.
(CVD) events, myocardial infarction, stroke, and CVD mortality, and these results were independent of baseline nutritional status. A prospective cohort study of middle-aged and elderly women also indicated no effect of MVM use for the same CVD outcomes in PHS II. The SU.VI.MAX Study, a clinical trial of antioxidative MVMs in adults, found no effect on incidence of ischaemic CVD and high-dose MVMs did not reduce CVD events. Meta-analysis of these and other studies (n=18) found no improvement in CVD outcomes in the general population. Based on these studies, the US Preventative Services Task Force does not recommend MVM use for the prevention of CVD.

Data on the effect of MVM consumption on cognitive function in adults are also inconclusive. While results from PHS II found that long-term use of daily MVs did not provide cognitive benefits in men, a meta-analysis on 10 studies concluded that MVs selectively enhanced free recall memory but no other cognitive functions. Intriguingly, 9 weeks of MVM use appears to improve multitasking and cognitive function during fatigue in women. With regard to cancer, PHS II demonstrated moderately reduced all-cancer risk in men consuming MVs, while data from the Women’s Health Initiative Clinical Trials revealed no association. Some studies even link MVM use with increased cancer risk—a prospective cohort study of Swedish women found increased breast cancer risk associated with MVM use.

The association of MVM use with all-cause mortality, like CVD, is null. While data from the Multiethnic Cohort Study cohort study indicated no association between MVM use and all-cause mortality, the Cancer Prevention Study (II) reported a 5% higher risk of all-cause death among men using MVs and the Iowa Women’s Health Study identified an association between MVM use and increased total mortality risk. A meta-analysis of these and other randomised trials (n=21) demonstrated no effect of MVM use on mortality risk.

While numerous reports on MVM consumption establish the lack of broad-spectrum, clinically measurable health benefits, the determinants of widespread MVM use by the general population are not well understood. The majority (52%) of MVM users report using MVMs in an effort to prevent disease, which is even more puzzling in light of the paucity of randomised and observation data showing a positive health benefit of MVMs. Understanding the determinants of widespread MVM use has significant medical and financial consequences. Moreover, it is unclear whether MVM users are different from non-users, simply believe they are healthier. To address this question, we used data from the 2012 National Health Interview Survey (NHIS), which included a complementary and alternative (CAM) questionnaire comprising 21,603 participants across the USA.

METHODS

Data source
All data were obtained from the 2012 National Health Interview Survey (NHIS), a nationally representative health survey conducted annually among civilian and non-institutionalised US participants by the Centers for Disease Control (CDC). All data were publicly available and did not require institutional review board approval. The 2012 NHIS was composed of a core questionnaire on health information administered to each selected household member. A randomly selected adult in each household was administered a more detailed health survey which included questions on access to care, specific health conditions and use of CAM (2012 only). In 2012, 77.6% of households completed the survey and 79.7% of adults selected completed the detailed survey.

Health status and health outcome measures
We obtained data on adults (age ≥18 years) derived from the Sample Adult Component who also participated in the Adult CAM File. This file surveys use of alternative medicines and therapies including daily MVM consumption, yoga and meditation. Consistent with previous NHIS studies, we considered five psychological, physical, and functional health outcomes from questions in the Sample Adult Component: (1) self-rated health status (poor/fair vs excellent/very good/good), (2) needing help with routine needs such as eating (yes or no), (3) history of 10 chronic diseases (cancer, hypertension, coronary heart disease, stroke, chronic obstructive pulmonary disease, asthma, diabetes, arthritis, hepatitis, and weak/failing kidneys), (4) presence of 19 health conditions in the past 12 months (digestive, skin, and other allergy, acid reflux, hay fever, chest cold, nausea and vomiting, sore throat, infectious disease, recurring headache, memory loss, neurological problems, sprains, and abdominal, dental, muscle/bone, chronic, and skin pain), and (5) Kessler 6-Item (K6) Psychological Distress Scale score to measure non-specific psychological distress in the past month. Participants who refused to answer or did not know the answers to at least one of these questions were excluded from the study. Participants were classified as MVM users or non-users from their response to the question ‘During the past 12 months, did take multi-vitamins or multi-minerals?’ in the Adult CAM File. Participants who refused to answer or did not know their MVM use in the past 12 months were excluded from analyses.

Statistical analysis
For each outcome, the association between MVM use in the past year and health outcome was estimated using a logistic regression model adjusting for age, sex, race, region, education, income, employment status, health insurance status, presence of child in household, marital status, unmet medical care due to cost in the past year, and not seeing a health professional in office in the past 2 weeks. Multinomial logistic regression was used for outcomes with more than two levels (eg, number
of chronic diseases, number of diseases in the past 12 months, Kessler-6 Item score). Binary logistic regression was used for outcomes with two levels (self-reported health and needing help with daily routines such as eating). Standard errors were estimated using weights provided by NHIS to account for the complex survey design and to produce nationally representative estimates. A multiple imputation strategy was used to estimate income in cases of missing responses to income as recommended by the National Centre for Health Statistics. All analyses were conducted using R (v3.5.1). p values were adjusted for multiple comparisons using a Benjamini-Hochberg procedure with false discovery rate (FDR) less than 0.01 deemed significant.

Stratified analyses were conducted in age (18–44, 45–64 and 65+years), race (white and non-white), sex (female and male), family income (<100%, 100–199%, 200–299%, 300–399% and 400% relative to the federal poverty level), and education level (did not graduate high school, high school graduate, college graduate or higher) groups to assess the association between MVM use and self-reported health in sociodemographic subgroups. In addition to stratified analyses, statistical interaction effects between MVM use and demographic variables (age, race, sex, family income, and education) on self-reported health were assessed using a multivariate regression model.

Patients and public involvement
Patients and the public were not involved in this study, including data collection, analysis and interpretation.

RESULTS
Study cohort characteristics
Sociodemographic differences between MVM users and non-users are presented in table 1. Our study included 4933 MVM users and 16670 non-users (table 1). As previously reported in data from the 2007–2010 and 2010–2014 National Health and Nutrition Examination Surveys (NHANES) compared with non-users, MVM users were significantly older, earned more income, were more likely to be female, more likely to be a college graduate, more likely to be married, and more likely to have health insurance. Unlike in previous studies, compared with MVM non-users, MVM users were less likely to be unemployed, to have a minor child in their household, and not to have an office visit for healthcare in the past 2 weeks (table 1). We observed no significant differences in percentage of non-English-speaking interviews and percentage having foregone medical care due to cost in the past year between MVM users and non-users (table 1).

Association between MVM usage and health status and health outcomes
Differences in health status and health outcomes between MVM users and non-users are displayed in table 1. Our study included 4933 MVM users and 16670 non-users (table 1). As previously reported in data from the 2007–2010 and 2010–2014 National Health and Nutrition Examination Surveys (NHANES) compared with non-users, MVM users were significantly older, earned more income, were more likely to be female, more likely to be a college graduate, more likely to be married, and more likely to have health insurance. Unlike in previous studies, compared with MVM non-users, MVM users were less likely to be unemployed, to have a minor child in their household, and not to have an office visit for healthcare in the past 2 weeks (table 1). We observed no significant differences in percentage of non-English-speaking interviews and percentage having foregone medical care due to cost in the past year between MVM users and non-users (table 1).

Stratified analyses: association between MVM usage and self-reported overall health in sociodemographic subgroups
Table 3 reports the association between MVM usage and self-reported overall health in age, race, sex, income and education stratified subgroups (table 3). MVM use was associated with better self-reported health in the group aged 18–44 years (OR 1.26, 95% CI 1.00 to 1.61) and 45–64 years (OR 1.30, 95% CI 1.08 to 1.57) and near significant among respondents aged 65 years and older (OR 1.20, 95% CI 0.95 to 1.52, FDR-adjusted p value=0.06) (table 3). MVM use was associated with better self-reported health among white (OR 1.34, 95% CI 1.07 to 1.67) and non-white (OR 1.26, 95% CI 1.09 to 1.45) respondents (table 3). MVM use was associated with better self-reported health in both male (OR 1.33, 95% CI 1.10 to 1.63) and female (OR 1.22, 95% CI 1.05 to 1.41) respondents (table 3). Interestingly, MVM use was associated with better self-reported health in families with income less than 100% of the federal poverty level (FPL) (OR 1.42, 95% CI 1.12 to 1.80), 100–199% FPL (OR 1.37, 95% CI 1.10 to 1.69) and 200–299% FPL (OR 1.32, 95% CI 1.01 to 1.72), but not in families whose income was 300–399% FPL (OR 1.32, 95% CI 0.88 to 1.98) or 400% FPL or higher (OR 1.15, 95% CI 0.85 to 1.56) (table 3). MVM use was associated with better self-reported health in all education subgroups analysed, including respondents who did not complete high school (OR 1.38, 95% CI 1.06 to 1.81), high school graduates (OR 1.21, 95% CI 1.04 to 1.41) and college graduates (OR 1.37, 95% CI 1.00 to 1.88) (table 3). All stratified analyses were conducted after adjusting for the potential confounding effects of age, sex, race, region, education, income, employment status, health insurance status, presence of child in household, marital status, unmet medical care due to cost in the past year, and not seeing a health professional in office in the past 2 weeks. The variable of stratification was not included as a covariate.

Statistical interaction effects between MVM use and demographic variables (age, race, family income and education) on self-reported overall health were assessed through a multivariate regression model in online supplemental table S1. We observed no significant association between MVM use and age, MVM use and race, MVM use and family income, and MVM use and education on self-reported overall income (online supplemental table S1).
Table 1  Characteristics of American Adults by Multivitamin and Multimineral Supplement (MVM) Usage

| Characteristic                                      | MVM non-users (n=4933*) | MVM users (n=16670*) | FDR-adjusted p value† |
|-----------------------------------------------------|-------------------------|----------------------|-----------------------|
| Weighted sample, %                                  | 22.4 (21.8 to 23.0)     | 77.6 (76.9 to 78.0)  |                       |
| Age, % (95% CI‡)                                     |                         |                      |                       |
| Mean age, years (95% CI)                            |                         |                      |                       |
| 18–27                                               | 48.1 (47.4 to 48.7)     | 49.7 (49.3 to 50.2)  | <0.001                |
| 28–37                                               | 14.9 (13.8 to 16.2)     | 13.1 (12.2 to 14.1)  |                       |
| 38–47                                               | 16.6 (15.4 to 18.0)     | 16.9 (16.2 to 17.7)  |                       |
| 48–57                                               | 17.4 (16.3 to 18.6)     | 15.3 (14.6 to 15.9)  |                       |
| 58–67                                               | 17.7 (16.4 to 19.0)     | 17.6 (16.9 to 18.3)  |                       |
| 68–80                                               | 14.3 (13.2 to 15.5)     | 15.4 (14.8 to 16.1)  |                       |
| >80                                                 | 10.1 (9.2 to 11.1)      | 12.8 (12.1 to 13.5)  |                       |
|                                                      | 5.9 (5.1 to 6.8)        | 6.2 (5.7 to 6.7)     |                       |
| Race, % (95% CI‡)                                   |                         |                      |                       |
| White only                                          | 82.2 (81.0 to 83.3)     | 82.9 (82.1 to 83.6)  | <0.001                |
| Black/African American only                         | 11.4 (10.4 to 12.5)     | 10.4 (9.9 to 11.0)   |                       |
| American Indian/Alaskan Native only                 | 1.1 (0.8 to 1.4)        | 0.6 (0.5 to 0.8)     |                       |
| Asian only                                          | 3.5 (3.1 to 4.0)        | 4.3 (3.9 to 4.6)     |                       |
| Multiple race                                       | 1.8 (1.5 to 2.2)        | 1.9 (1.6 to 2.1)     |                       |
| Female, % (95% CI‡)                                 | 54.1 (52.6 to 55.6)     | 59.1 (58.2 to 60.1)  | <0.001                |
| Family income relative to federal poverty level, % (95% CI‡) |                   |                      |                       |
| <100                                                | 16.9 (15.3 to 18.4)     | 12.4 (11.5 to 13.3)  | <0.001                |
| 100–199                                             | 19.7 (18.2 to 21.2)     | 17.9 (17.1 to 18.8)  |                       |
| 200–299                                             | 17.3 (15.8 to 18.7)     | 17.0 (16.2 to 17.8)  |                       |
| 300–399                                             | 12.8 (11.4 to 14.2)     | 13.4 (12.6 to 14.1)  |                       |
| ≥400                                                | 33.4 (31.1 to 35.6)     | 39.4 (37.9 to 40.9)  |                       |
| Education status, % (95% CI‡)                       |                         |                      |                       |
| Did not graduate high school                        | 11.7 (10.7 to 12.8)     | 9.6 (9.0 to 10.1)    | <0.001                |
| Grade 12 or GED                                     | 26.6 (24.8 to 28.5)     | 22.4 (21.4 to 23.4)  |                       |
| Some college, no degree                             | 22.1 (20.5 to 23.8)     | 21.2 (20.1 to 22.4)  |                       |
| Associates degree                                   | 10.8 (9.7 to 11.9)      | 12.0 (11.4 to 12.6)  |                       |
| College graduate or higher                          | 28.7 (26.7 to 30.7)     | 34.7 (33.3 to 36.2)  |                       |
| Relationship status, % (95% CI‡)                    |                         |                      |                       |
| Married or living with partner                       | 49.0 (46.4 to 51.7)     | 51.0 (49.4 to 52.7)  | <0.001                |
| Separated, divorced or widowed                      | 26.6 (25.0 to 28.3)     | 26.7 (25.6 to 27.8)  |                       |
| Never married                                       | 24.3 (22.5 to 26.1)     | 22.3 (21.0 to 23.5)  |                       |
| Employment status, % (95% CI‡)                      |                         |                      |                       |
| Employed                                            | 58.1 (55.2 to 60.9)     | 58.6 (56.7 to 60.5)  | 0.05                  |
| Unemployed, looking for work                        | 6.1 (5.2 to 7.0)        | 5.2 (4.8 to 5.6)     |                       |
| Not in labour force                                 | 35.8 (33.7 to 37.9)     | 36.2 (34.8 to 37.6)  |                       |
| Minor child in household, % (95% CI‡)               | 30.4 (28.8 to 32.0)     | 26.5 (25.5 to 27.3)  | <0.001                |
| Non-English-speaking interview, % (95% CI‡)         | 3.6 (3.1 to 4.1)        | 3.5 (3.1 to 3.8)     | 0.66                  |
| Has health insurance, % (95% CI‡)                   | 84.3 (83.1 to 85.4)     | 87.4 (86.9 to 88.0)  | <0.001                |
| No office visit for healthcare in the past 2 weeks, % (95% CI‡) | 79.8 (78.6 to 81.0) | 76.4 (75.7 to 77.1)  | <0.001                |
| Unmet medical care due to cost in the past year, % (95% CI‡) | 9.4 (8.5 to 10.3) | 8.7 (8.3 to 9.2)     | 0.19                  |

*Unweighted sample size.
†FDR-adjusted p value was computed using the Benjamini-Hochberg procedure; p values were computed using a two-sample t-test or χ² test for independence.
‡All confidence intervals were computed based on a Rao-Scott-scaled χ² distribution for the loglikelihood from a binomial distribution using the Survey package in R.
§GED indicates completion of General Educational Development and provides certification of high-school level credentials in the United States.
DISCUSSION

This present study is the first to simultaneously analyse the association between MVM use and both self-reported health and clinical health outcomes. In this work, we found that MVM users self-report 30% better overall health than non-users despite no clinically assessed differences in health. Our finding that MVM users and non-users do not differ in various psychological, physical and functional outcomes corroborates previous reports that MVMs do not improve overall health in the general adult population.5–22 Our stratified analysis revealed that MVM use is associated with better self-reported overall health across all race, sex and education groups, and in individuals under 65 and with family incomes below 300% FPL. The lack of association between MVM usage and self-reported health in individuals with family income greater than 300% FPL may be related to sample size and should be replicated in a follow-up study. Taken together, these findings help elucidate explanations underlying widespread MVM usage despite no generalised clinical benefits.

The results here suggest two potential explanations underlying widespread MVM consumption in the absence of clinically measurable benefits: MVM users believe in the efficacy of MVMs by harbouring a positive expectation regarding the health benefits of MVMs; and MVM users believe in the efficacy of MVMs by harbouring a positive expectation regarding the health benefits of MVMs; and MVM

| Table 2  | Association between MVM usage and health status |
|----------|-----------------------------------------------|
| Characteristic | MVM non-users | MVM users | Adjusted effect of MVM usage, β or OR (95% CI)* | FDR-adjusted p value† |
| Self-rated overall health as excellent, very good or good, % (95% CI‡) | 84.9 (83.8 to 86.0) | 88.3 (87.7 to 88.9) | OR=1.3 (1.2 to 1.5) | <0.001 |
| Needs help with ADLs, % (95% CI‡) | 5.6 (4.8 to 6.3) | 4.8 (4.4 to 5.2) | OR=0.86 (0.7 to 1.04) | 0.07 |
| History of chronic conditions**, % (95% CI‡) | | | | |
| Mean number of chronic conditions | 1.07 (1.03 to 1.11) | 1.09 (1.06 to 1.11) | β=0.03 (−0.07 to 0.007) | 0.07 |
| No chronic conditions | 44.4 (42.0 to 46.8) | 43.0 (41.4 to 44.5) | | |
| One chronic condition | 26.3 (24.5 to 28.2) | 26.4 (25.4 to 27.5) | | |
| Multiple chronic conditions | 28.4 (26.7 to 30.0) | 29.7 (28.6 to 30.7) | | |
| Health conditions in past year§, (95% CI‡) | | | | |
| Mean number of present conditions | 2.8 (2.7 to 2.9) | 2.7 (2.7 to 2.8) | β=−0.06 (−0.2 to 0.02) | 0.08 |
| 0–5 present conditions | 84.7 (81.3 to 88.1) | 85.2 (83.0 to 87.6) | | |
| 6–10 present conditions | 12.7 (11.6 to 13.8) | 12.4 (11.7 to 13.0) | | |
| ≥10 present conditions | 1.5 (1.1 to 1.9) | 1.4 (1.2 to 1.6) | | |
| Kessler 6-Item score, % (95% CI‡) | | | | |
| Mean Kessler score | 2.5 (2.4 to 2.6) | 2.3 (2.3 to 2.4) | β=−0.08 (−0.2 to 0.04) | 0.13 |
| No impairment | 80.9 (77.4 to 84.4) | 82.3 (80.0 to 84.6) | | |
| Moderate Impairment | 15.4 (14.2 to 16.6) | 14.8 (14.1 to 15.5) | | |
| Severe Impairment | 3.7 (3.1 to 4.2) | 2.9 (2.6 to 3.2) | | |

*Estimates were produced after adjusting for age, sex, race, region, education level, income, employment status, health insurance status, presence of child in household, marital status, unmet medical care due to cost in the past year, and not seeing a health professional in office in the past 2 weeks using a multivariate regression model.
†FDR-adjusted p values were computed using the Benjamini-Hochberg procedure.
‡All confidence intervals were computed based on a Rao-Scott-scaled χ² distribution for the loglikelihood from a binomial distribution using the Survey package in R.
§19 health conditions in the past 12 months included: respiratory, digestive, skin, and other allergy, acid reflux, hay fever, chest cold, nausea and vomiting, sore throat, infectious disease, recurring headache, memory loss, neurological problems, sprains, and abdominal, dental, muscle/bone, chronic, and skin pain.
¶P value was defined using a multivariate regression model controlling for age, sex, race, region, education level, income, employment status, health insurance status, presence of child in household, marital status, unmet medical care due to cost in the past year, and not seeing a health professional in office in the past 2 weeks.
**Ten chronic diseases included: cancer, hypertension, coronary heart disease, stroke, chronic obstructive pulmonary disease, asthma, diabetes, arthritis, hepatitis, and weak/failing kidneys.
ADL, activities of daily living; FDR, false discovery rate; MVM, multivitamin and multiminer.
users intrinsically harbour a more positive outlook on their personal health regardless of MVM usage. A growing body of evidence suggests that positive expectation influences treatment outcomes for diseases including heart disease,32–35 cancer,36 37 musculoskeletal disorders,38 39 injuries40 41 and obesity.42–44 Under a positive expectation model, MVM users are more likely to harbour a positive expectation regarding the clinical efficacy of MVMs and thus are more likely to self-report as having excellent or good overall health. In the case of MVM usage, it is interesting that the presence of positive expectations did not influence clinically measurable health outcomes, unlike in other treatments. The effect of positive expectations in the MVM user community is made even stronger when one considers that the majority of MVM and supplements are sold to the so-called ‘worried-well’ population45 who may assign greater weight to the purported health benefits of dietary supplements and alternative therapies. It is possible that members of this population are more susceptible to positive expectations and may thereby continue to use MVMs in the absence of clinical benefits.

The second mechanism, in which MVM users intrinsically harbour greater positive views about their health, may be explained in part by certain combinations of sociodemographic determinants that influence self-reported health. While age, sex, income, education and location of residence have been previously shown to affect self-reported health in diverse populations,46–48 combinations of other characteristics may also cause MVM users to harbour intrinsically more positive views regarding their health in the absence of clinical differences. Further research is necessary to elucidate these characteristics.

Our results are consistent with existing work from two studies: the first being a 2013 study involving 11,956

| Table 3 Association between MVM usage and self-reported overall health in sociodemographic subgroups |
|---------------------------------------------|---------------------------------------------|---------------------------------------------|---------------------------------------------|---------------------------------------------|
| Group                                      | Self-rated overall health as excellent, very good or good, % (95% CI)*, MVM Non-Users | Self-rated overall health as excellent, very good or good, % (95% CI)*, MVM Users | Adjusted effect on self-reported health, OR (95% CI)† | FDR adjusted p value‡ |
|---------------------------------------------|---------------------------------------------|---------------------------------------------|---------------------------------------------|---------------------------------------------|
| Age, years                                 |---------------------------------------------|---------------------------------------------|---------------------------------------------|---------------------------------------------|
| 18–44                                      | 92.3 (91.1 to 93.5)                         | 94.2 (93.6 to 94.8)                         | 1.3 (1.0 to 1.6)                            | 0.03                                       |
| 45–64                                      | 79.9 (77.8 to 82.1)                         | 85.3 (84.2 to 86.4)                         | 1.3 (1.1 to 1.6)                            | 0.009                                      |
| ≥65                                        | 77.2 (73.8 to 80.5)                         | 82.0 (80.6 to 83.4)                         | 1.2 (1.0 to 1.5)                            | 0.06                                       |
| Race                                        |---------------------------------------------|---------------------------------------------|---------------------------------------------|---------------------------------------------|
| White                                      | 85.9 (84.7 to 87.2)                         | 89.1 (88.5 to 89.7)                         | 1.3 (1.1 to 1.7)                            | 0.009                                      |
| Non-white                                  | 80.0 (77.2 to 82.7)                         | 84.2 (82.8 to 85.6)                         | 1.3 (1.1 to 1.5)                            | 0.007                                      |
| Sex                                         |---------------------------------------------|---------------------------------------------|---------------------------------------------|---------------------------------------------|
| Female                                     | 84.0 (82.5 to 85.4)                         | 88.1 (87.4 to 88.9)                         | 1.2 (1.1 to 1.4)                            | 0.009                                      |
| Male                                        | 85.9 (84.2 to 87.7)                         | 88.4 (87.5 to 89.3)                         | 1.3 (1.1 to 1.6)                            | 0.009                                      |
| Family income relative to federal poverty level, % (95% CI) |---------------------------------------------|---------------------------------------------|---------------------------------------------|---------------------------------------------|
| <100                                       | 71.7 (68.0 to 75.4)                         | 75.6 (73.1 to 78.1)                         | 1.4 (1.1 to 1.8)                            | 0.007                                      |
| 100–199                                     | 76.4 (73.6 to 79.2)                         | 80.7 (79.0 to 82.4)                         | 1.4 (1.1 to 1.7)                            | 0.007                                      |
| 200–299                                     | 84.8 (82.1 to 87.5)                         | 87.3 (85.9 to 88.6)                         | 1.3 (1.0 to 1.7)                            | 0.04                                       |
| 300–399                                     | 89.6 (86.4 to 92.7)                         | 91.0 (89.6 to 92.4)                         | 1.3 (0.9 to 2.0)                            | 0.15                                       |
| ≥400                                       | 94.8 (93.5 to 96.1)                         | 95.2 (94.6 to 95.8)                         | 1.1 (0.8 to 1.6)                            | 0.23                                       |
| Education                                   |---------------------------------------------|---------------------------------------------|---------------------------------------------|---------------------------------------------|
| Did not graduate high school                | 67.2 (63.1 to 71.3)                         | 71.9 (69.7 to 74.2)                         | 1.4 (1.1 to 1.9)                            | 0.01                                       |
| High school graduate                        | 84.1 (82.6 to 85.5)                         | 86.7 (85.9 to 87.4)                         | 1.2 (1.0 to 1.4)                            | 0.01                                       |
| College graduate or higher                  | 93.8 (92.4 to 95.1)                         | 95.3 (94.7 to 95.9)                         | 1.4 (1.0 to 1.9)                            | 0.03                                       |

*All confidence intervals were computed based on a Rao-Scott-scaled χ² distribution for the loglikelihood from a binomial distribution using the Survey package in R.
†Estimates were produced after adjusting for age, sex, race, region, education level, income, employment status, health insurance status, presence of child in household, marital status, unmet medical care due to cost in the past year, and not seeing a health professional in office in the past 2 weeks.
‡FDR-adjusted p values were computed using the Benjamini-Hochberg procedure; p value was defined using a multivariate regression model controlling for age, sex, race, region, education level, income, employment status, health insurance status, presence of child in household, marital status, unmet medical care due to cost in the past year, and not seeing a health professional in office in the past 2 weeks.
FDR, false discovery rate; MVM, multivitamin and multimineral.
adults from the 2007–2010 NHANES that demonstrated MVM users exhibit better self-reported health than non-users; and second, a 2014 study involving 5536 Coast Guard and military personnel which found that MVM users were significantly more likely to self-report their general health as excellent or good. While informative, these previous studies only focused on self-reported health as an outcome. In the present study, we considered self-reported health in addition to clinically measurable health outcomes. This is an important distinction establishing that MVM users experience better self-reported health in the absence of clinically measurable health improvement. Nevertheless, it is encouraging that our results are consistent across the NHANES, military and Coast Guard, and NHIS study cohorts, and robust to different statistical analysis methodologies.

Limitations of this study include the cross-sectional design, reliability of self-reported health within NHIS, income estimation using multiple imputation, indication bias and non-response bias. First, the cross-sectional study design prevents demonstration of a causal relationship between MVM use and self-reported health. The lack of longitudinal data available to assess changes in self-reported health before and after MVM supplementation prevents us from differentiating the two aforementioned explanations that may contribute to widespread MVM use. Second, self-reported health within the NHIS may inherently harbour reporting bias and residual confounding. In addition to reporting bias and residual confounding, a self-reported binary response to the question of whether one has taken MVMs in the past 12 months precludes any analysis of the dose-dependent effects of MVMs in our cohort. This is especially important considering some vitamins and minerals have known U-shaped associations with disease, where disease risk is elevated at both high and low vitamin and mineral levels. Further, use of both multivitamins and multiminerals were asked together as part of the same question in the NHIS questionnaire. This prevented us from analysing multivitamin and multimineral effects in isolation. Moreover, different MVM preparations can differ in their nutritional composition, quality, and bioavailability. Some individuals may take multiple MVMs whose constituents could interact with each other. Because the brand of multivitamin being taken was not asked of MVM users in NHIS, we could not identify differences in nutritional composition, quality, bioavailability and chemical interaction that may be driving the results in this study.

Third, despite being recommended by the NHIS, the multiple imputation technique used to calculate income in cases in which data were missing may generate estimation errors. Another limitation to the income-stratified results for self-reported overall health may stem from the inability to take account of income mobility. Interestingly, it has been previously demonstrated that while high incomes are associated with longer life expectancies, accounting for income mobility reduces the gap by approximately 50%.

A portion of our cohort may have been prescribed MVMs, specific vitamins or specific minerals for indications including micronutrient deficiency, pregnancy, iron deficiency anaemia, osteoporosis, Crohn’s disease and others, thereby contributing to indication bias. Previous estimates have suggested approximately 1% of physician office visits in the USA include a prescription or recommendation for MVMs. One can imagine a scenario in which MVM users and non-users are imbalanced in the proportion of medical cases that require MVM supplementation (ie, micronutrient deficiency or pregnancy). In such a scenario, it may falsely appear that MVM use is not associated with clinical benefits. In the present study, owing to a lack of information regarding the reason for taking MVMs, we were unable to fully account for indication bias present in our cohort.

In addition to indication bias, the NHIS, like other surveys, is known to suffer from non-response bias. For example, a previous study found that the 1990–2009 NHIS population had an approximately 14% lower mortality than the general population. Post hoc methods to address non-response bias include creating sample weights based on demographic variables and selection probabilities, as was used in the present study. However, survey weighting, while a standard practice, may not fully account for non-response bias, especially if the survey weights do not take into account common differences between survey responders and non-responders, such as smoking and alcohol use. As a result, non-response bias may limit the generalisability of our results to the broader population.

CONCLUSIONS

Using nationally representative survey data on health outcomes, our study reveals that MVM users self-report better overall health than non-users despite not exhibiting improved health by clinically measurable standards. Furthermore, we identify specific sociodemographic subgroups of MVM users that are more prone to this behaviour. The multibillion-dollar nature of the nutritional supplement industry means that understanding the determinants of widespread MVM use has significant medical and financial consequences. Our findings suggest that widespread use of multivitamins in adults may be a result of individuals’ positive expectation that multivitamin use leads to better health outcomes or a self-selection bias in which MVM users intrinsically harbour more positive views regarding their health.

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Health benefits of multivitamins/mineral supplements may be all in the minds of users

No measurable clinical improvements between them and non-users, research suggests

The health benefits of multivitamin/mineral supplements may be all in the minds of those who take them, prompted by positive expectations of effectiveness rather than hard evidence of that, suggests research published in the online journal *BMJ Open*.

Multivitamin/mineral supplements are widely used, with some estimates indicating that one in three Americans alone takes them regularly: the industry is worth billions of dollars.

Yet a succession of clinical trials has failed to identify clinically measurable health benefits associated with daily consumption among those without known vitamin or mineral deficiencies, say the researchers.

To explore this further, they compared self-reported and clinically measurable health outcomes among users and non-users of multivitamin/mineral supplements.

They drew on information supplied by a large, nationally representative sample of 21,603 US adults, all of whom had been additionally quizzed about their use of complementary medicine in the 2012 annual National Health Interview Survey.

The researchers compiled five psychological, physical, and functional health outcomes from the questions asked.

These were: subjective assessment of health; need for help with routine daily activities; history of 10 long term conditions, such as high blood pressure, asthma, diabetes, and arthritis; presence of 19 common health conditions in the preceding 12 months, including infections, memory loss, neurological and musculoskeletal problems; and degree of psychological distress, measured on a validated scale (K6).

Some 4933 respondents said they regularly took multivitamin/mineral supplements; 16670 respondents said they didn’t.

Regular users were significantly older and had higher household incomes than those who didn’t take multivitamin/mineral supplements. They were also more likely to be women, a college graduate, married, and to have health insurance.

Regular multivitamin/mineral supplement users reported 30% better overall health than those who didn’t take them. But there was no difference between those who did and didn’t take them in any of the five psychological, physical or functional health outcomes assessed.

Further in-depth analysis revealed that regularly taking multivitamin/mineral supplements was associated with better self-reported overall health across all race, sex, and education groups, as well as in the under 65s and those on low household incomes.

This is an observational study, and so can’t prove a causal relationship between use of multivitamin/mineral supplements and subjectively assessed health. What’s more, subjectively assessed health is not always reliable, and the researchers didn’t track changes in health before and after taking supplements over the long term.
Nevertheless, the lack of any difference in the health outcomes assessed is in line with other studies indicating that multivitamin/mineral supplements don't improve overall health in the general adult population, say the researchers.

They put forward two possible explanations for their findings: either regular users of these supplements simply believe they work and that they confer health benefits in the absence of hard evidence to that effect, or they are just inherently more positive about their personal health, irrespective of their supplement use.

A growing body of evidence implicates the power of positive thinking in the improvement of various health conditions, the researchers point out.

“The effect of positive expectations in [those who take multivitamin/mineral supplements] is made even stronger when one considers that the majority of [them] are sold to the so-called ‘worried-well,’” they say.

“The multibillion-dollar nature of the nutritional supplement industry means that understanding the determinants of widespread [multivitamin/mineral] use has significant medical and financial consequences,” they point out.
