

Funding and Publication of Research on Gun Violence and Other Leading Causes of Death

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Online Supplement

This [R Markdown document](#) contains detailed methods and annotated code associated with [Stark DE, Shah NH. Funding and Publication of Research on Gun Violence and Other Leading Causes of Death. JAMA. 2017;317\(1\)](#) and enables fully reproducible research.

The script `gun_violence_annotated_code.Rmd` should be run within the same working directory as the accompanying file, `Compressed Mortality, 2004-2014.txt`. Internet access is required, as the code will retrieve data from the MEDLINE and Federal RePORTER databases. Running this `.Rmd` script in RStudio will take approximately 4-6 hours to complete on a standard machine.

load required libraries

```
# load libraries
require(RCurl)
require(XML)
require(ggplot2)
require(scales)
library(pander)
require(dplyr)
require(RColorBrewer)
require(ggrepel)
require(SPARQL)
require(gridExtra)
```

CDC Mortality Rates

[CDC mortality statistics](#) were accessed from 2004 to 2014 (the most recent year available). Results were grouped by 'Injury Mechanism & All Other Leading Causes' and sorted by mortality rate. 13 nonspecific causes of death were excluded (see below) and the top 30 causes of death were retained for further analysis.

Nonspecific causes of death excluded from analysis:

All other diseases (Residual); Symptoms, signs and abnormal clinical and laboratory findings, not elsewhere classified; Other diseases of respiratory system; Other diseases of the circulatory system; In situ neoplasms, benign neoplasms and neoplasms of uncertain or unknown behavior; Certain conditions originating in the perinatal period; Congenital malformations, deformations and chromosomal abnormalities; Unspecified Injury; Other and unspecified infectious and parasitic diseases and their sequelae; Other disorders of circulatory system; Complications of medical and surgical care; Other specified,

classifiable Injury; Other specified, not elsewhere classified Injury

CDC-derived causes of death were manually mapped to their corresponding Medical Subject Heading (MeSH) term(s). Ambiguous mappings were resolved by inspecting ICD-10 codes associated with a particular cause of death.

The downloaded CDC file Compressed Mortality, 2004-2014.txt was annotated with 4 additional columns prior to importing for analysis:

- Remove Flag indicating nonspecific causes of death for removal
- MeSH.Terms Mapped term(s) corresponding to CDC-derived cause of death
- MeSH.IDs Corresponding MeSH Unique ID(s)
- Abbreviation Abbreviated term used for plots

Code to import CDC mortality data

```
# import CDC mortality data with manually mapped MeSH terms
mortality <- read.delim("Compressed Mortality, 2004-2014.txt", stringsAsFactors = FALSE)

# remove nondescript causes of death
mortality <- filter(mortality, Remove == FALSE)

# Create 'Cause' column defining injury versus non-injury
for (row in 1:nrow(mortality)) {
  if (substr(mortality$Injury.Mechanism...All.Other.Learning.Causes[row], start = 1,
    stop = 10) == "Non-Injury") {
    mortality$Cause[row] <- "Non-Injury"
  } else {
    mortality$Cause[row] <- "Injury"
  }
}

# convert multiple MeSH queries into list
convertList <- function(terms) {
  as.list(toupper(strsplit(terms, ";")[[1]]))
}
mortality$MeSH.Terms <- sapply(mortality$MeSH.Terms, convertList)
mortality$MeSH.IDs <- sapply(mortality$MeSH.IDs, convertList)
```

MEDLINE Publication Volume

For each cause of death, MEDLINE was queried for the total number of publications between 2004 and 2015 indexed with the corresponding MeSH term(s) including descendant terms (terms subsumed under a parent term within the MeSH hierarchy).

This was performed using the MEDLINE E-utilities API and the code below.

Code to import MEDLINE publication data

```

# Return total number of articles for each set of MeSH queries

# Generate PubMed query
mortality$PubMed.Query <- sapply(mortality$MeSH.Terms, paste, "[mesh]", sep = "",
  collapse = " OR ")

getPubMedTrend <- function(query, minYear = 2004, maxYear = 2015) {
  # Retrieves PubMed trend (counts results by year). Args: query: <string>
  # Search query minYear: <int> minimum year to return maxYear: <int> maximum
  # year to return Returns: A table containing year, count

  # PubMed EUtils URL for retrieving search results counts
  pubmed <- "https://eutils.ncbi.nlm.nih.gov/entrez/eutils/esearch.fcgi?db=pubmed;rettype=count;term="
  # encode query as a URL
  query <- URLencode(query)
  curl <- getCurlHandle()
  output <- data.frame(NULL)
  # retrieve counts for each year in range
  for (i in minYear:maxYear) {
    query_year <- paste(query, "+AND+", i, "%5Bppdat%5D", sep = "")
    result <- getURL(paste(pubmed, query_year, sep = ""), curl = curl)
    result <- xmlTreeParse(result, asText = TRUE)
    count <- as.numeric(xmlValue(result[["doc"]][["eSearchResult"]][["Count"]]))
    output <- rbind(output, data.frame(Year = i, Count = count))
  }

  return(output)
}

# For each cause of death, run PubMed query and sum total results over
# 2004-2015
mortality$Publications <- NA
x = 1
for (query in mortality$PubMed.Query) {
  mortality$Publications[x] <- colSums(getPubMedTrend(query, 2004, 2015))[2]
  x = x + 1
}

```

Federal RePORTER Funding Data

Research funding data from 2004 to 2015 (all years available) was accessed from [Federal RePORTER](#), a database of projects funded by U.S. federal agencies. Projects are indexed using the computerized Research, Condition, and Disease Categorization system derived in part from MeSH. For each cause of death, Federal RePORTER was queried for the total funding awarded to projects containing corresponding MeSH terms, including descendant terms.

Code to import Federal RePORTER funding data

```

# Import Federal ExPORTER data

# Code to download zip csv files from url

```

```

temp <- tempfile()
download.file("https://federalreporter.nih.gov/FileDownload/DownloadFile?fileToDownload=FedRePORTER_PRJ_
temp)
data <- read.csv(unz(temp, "FedRePORTER_PRJ_C_FY2004.csv"), stringsAsFactors = FALSE)
unlink(temp)

FedReporter <- NA
for (year in 2004:2015) {
  temp <- tempfile()
  download.file(paste("https://federalreporter.nih.gov/FileDownload/DownloadFile?fileToDownload=FedRe
    year, ".zip", sep = ""), temp)
  data <- read.csv(unz(temp, paste("FedRePORTER_PRJ_C_FY", year, ".csv", sep = "")),
    stringsAsFactors = FALSE, header = FALSE, skip = 1)
  unlink(temp)
  FedReporter <- rbind(FedReporter, data)
}
colnames(FedReporter) <- c("SM_Application_ID", "Project_Terms", "Project_Title",
  "Department", "Agency", "IC_Center", "Project_Number", "Project_Start_Date",
  "Project_End_Date", "Contact_PI_Project_Leader", "Other_PIs", "Congressional_District",
  "DUNS_Number", "Organization_Name", "Organization_City", "Organization_State",
  "Organization_Zip", "Organization_Country", "Budget_Start_Date", "Budget_End_Date",
  "CFDA_Code", "FY", "FY_Total_Cost", "FY_Total_Cost_Sub_Projects")

# convert project terms to list and all caps
FedReporter$Project_Terms <- sapply(FedReporter$Project_Terms, convertList)
# convert funding NAs to zeros
FedReporter$FY_Total_Cost[is.na(FedReporter$FY_Total_Cost)] <- 0

```

MeSH term expansion

In order to ensure complete coverage of search terms, MeSH terms were expanded to include all descendant terms (MEDLINE does this automatically in its queries but Federal RePORTER does not.) The [MeSH SPARQL endpoint](#) was used to perform MeSH term expansion.

```

# For each MeSH query (or set of MeSH queries) return list of descendant
# queries

stripExtra <- function(term) {
  substr(term, 2, nchar(term) - 4)
}

getChildren <- function(term) {
  endpoint <- "https://id.nlm.nih.gov/mesh/sparql"

  # Query to retrieve all synonym terms of the input term
  query <- paste("PREFIX mesh: <http://id.nlm.nih.gov/mesh/>
PREFIX mesh2015: <http://id.nlm.nih.gov/mesh/2015/>
PREFIX mesh2016: <http://id.nlm.nih.gov/mesh/2016/>
PREFIX meshv: <http://id.nlm.nih.gov/mesh/vocab#>
PREFIX rdfs: <http://www.w3.org/2000/01/rdf-schema#>

SELECT DISTINCT ?labelA
FROM <http://id.nlm.nih.gov/mesh>

```

```

WHERE {{
    mesh:",
term, " rdfs:label ?labelA .
    } UNION
    {
    mesh:",
term, " meshv:treeNumber ?treeNum .
    ?childTreeNum meshv:parentTreeNumber+ ?treeNum .
    ?descriptorA meshv:treeNumber ?childTreeNum .
    ?descriptorA rdfs:label ?labelA .
    }}" ,

    sep = "")
df <- SPARQL(endpoint, query, extra = "format=HTML&inference=TRUE")$results
colnames(df) <- NULL
return(toupper(sapply(df, stripExtra)))
}

for (row in 1:nrow(mortality)) {
  mortality$MeSH.Children[row] <- paste(unlist(sapply(mortality$MeSH.IDs[[row]],
    getChildren)), collapse = ";")
}

mortality$MeSH.Children <- sapply(mortality$MeSH.Children, convertList)

# function to invert strings with commas
invertCommas <- function(term) {
  s1 <- (strsplit(term, ", "))[1] # split at commas
  output <- paste(rev(s1), collapse = " ") # reverse and collapse
  return(output)
}

for (causeNum in 1:length(mortality$MeSH.Children)) {
  for (termNum in 1:length(mortality$MeSH.Children[[causeNum]])) {
    mortality$MeSH.Children[[causeNum]][[termNum]] <- invertCommas(mortality$MeSH.Children[[causeNum]][[termNum]])
  }
}

# For each bundled set of children queries, search Federal RePORTER and
# return total funding

getFunding <- function(terms) {
  funding = 0
  projects = 0
  row = 1
  for (row in 1:nrow(FedReporter)) {
    if (length(intersect(terms, FedReporter$Project_Terms[row][[1]])) !=
      0) {
      funding = funding + FedReporter$FY_Total_Cost[row]
      projects = projects + 1
    }
  }
  return(c(funding, projects))
}

```

```

}

mortality$Total.Funding <- NA
mortality$Total.Projects <- NA
for (row in 1:nrow(mortality)) {
  result <- getFunding(mortality$MeSH.Children[[row]])
  mortality$Total.Funding[row] <- result[1]
  mortality$Total.Projects[row] <- result[2]
}

data <- select(mortality, Cause, Abbreviation, Crude.Rate, Publications, Total.Funding,
  Total.Projects)

# Sort by mortality rate and filter top 30 causes of death for inclusion
data <- slice(arrange(data, desc(Crude.Rate)), 1:30)

write.csv(x = data, file = "gun_violence_data_frame.csv")

```

Results

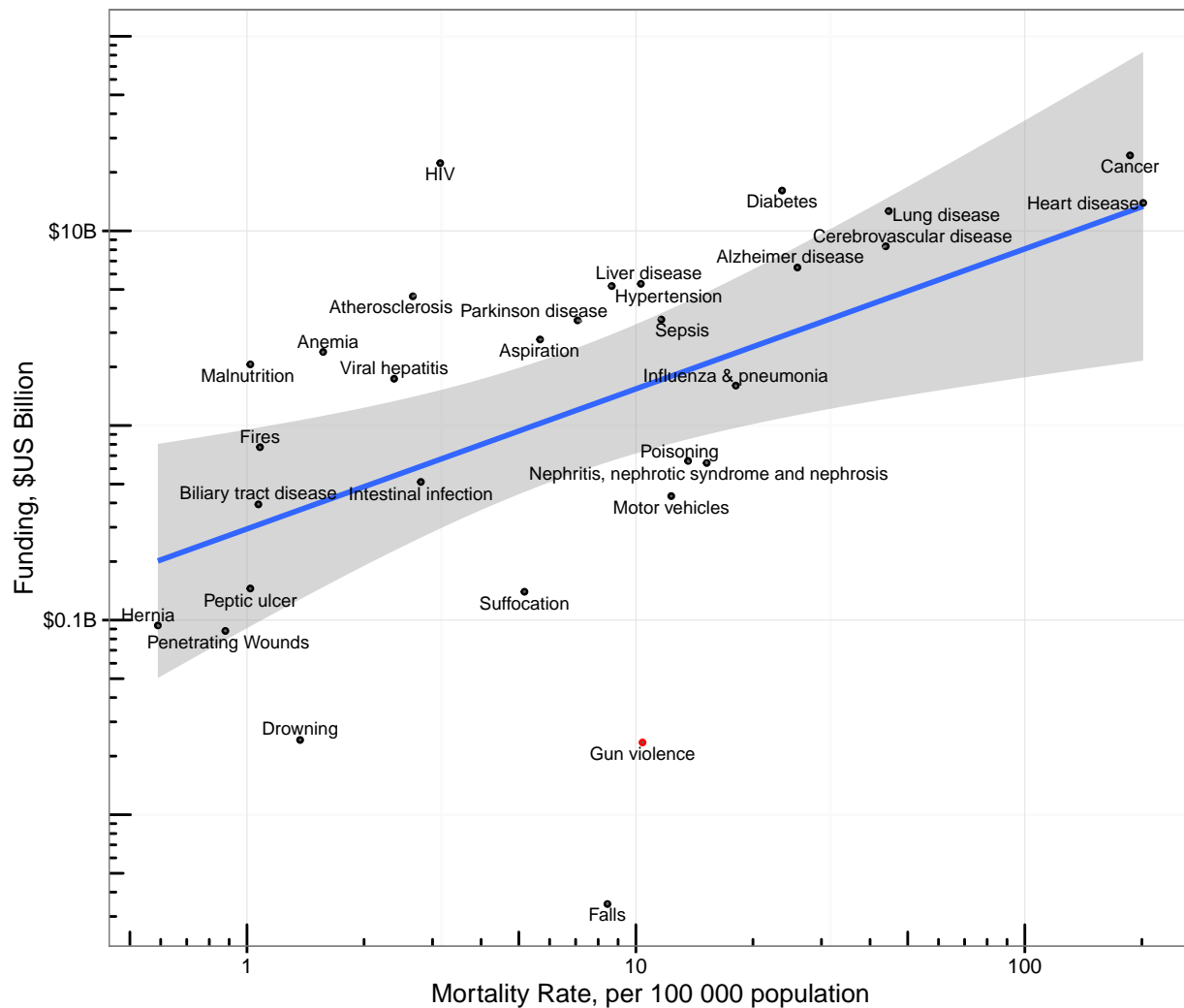
Code to plot funding and publication volume for leading causes of death

```

# formatting for funding axis labels
funding_format <- function(x) {
  return(paste("$", x, "B", sep = ""))
}

# plot log(mortality) x log(funding)
Fig1A_funding <- ggplot(data, aes(x = Crude.Rate, y = (Total.Funding/1e+09))) +
  stat_smooth(method = "lm") + geom_point(size = 0.75, color = as.numeric(data$Abbreviation ==
  "Gun violence") + 1) + scale_color_brewer(type = "qual", palette = "Set1") +
  geom_text_repel(aes(label = Abbreviation), size = 2.4, segment.size = 0.5,
    box.padding = unit(0.1, "lines")) + scale_y_log10(labels = funding_format) +
  scale_x_log10() + annotation_logticks() + theme_bw(base_size = 10) + labs(y = "Funding, $US Billion",
  labs(x = "Mortality Rate, per 100 000 population")
Fig1A_funding

```



```
# plot log(mortality) x log(publications)
Fig1B_publications <- ggplot(data, aes(x = Crude.Rate, y = (Publications/1000))) +
  stat_smooth(method = "lm") + geom_point(size = 0.75, color = as.numeric(data$Abbreviation ==
    "Gun violence") + 1) + scale_color_brewer(type = "qual", palette = "Set1") +
  geom_text_repel(aes(label = Abbreviation), size = 2.4, segment.size = 0.5,
    box.padding = unit(0.1, "lines")) + scale_y_log10() + scale_x_log10() +
  annotation_logticks() + theme_bw(base_size = 10) + labs(y = "Publications, in Thousands") +
  labs(x = "Mortality Rate, per 100 000 population")
Fig1B_publications
```

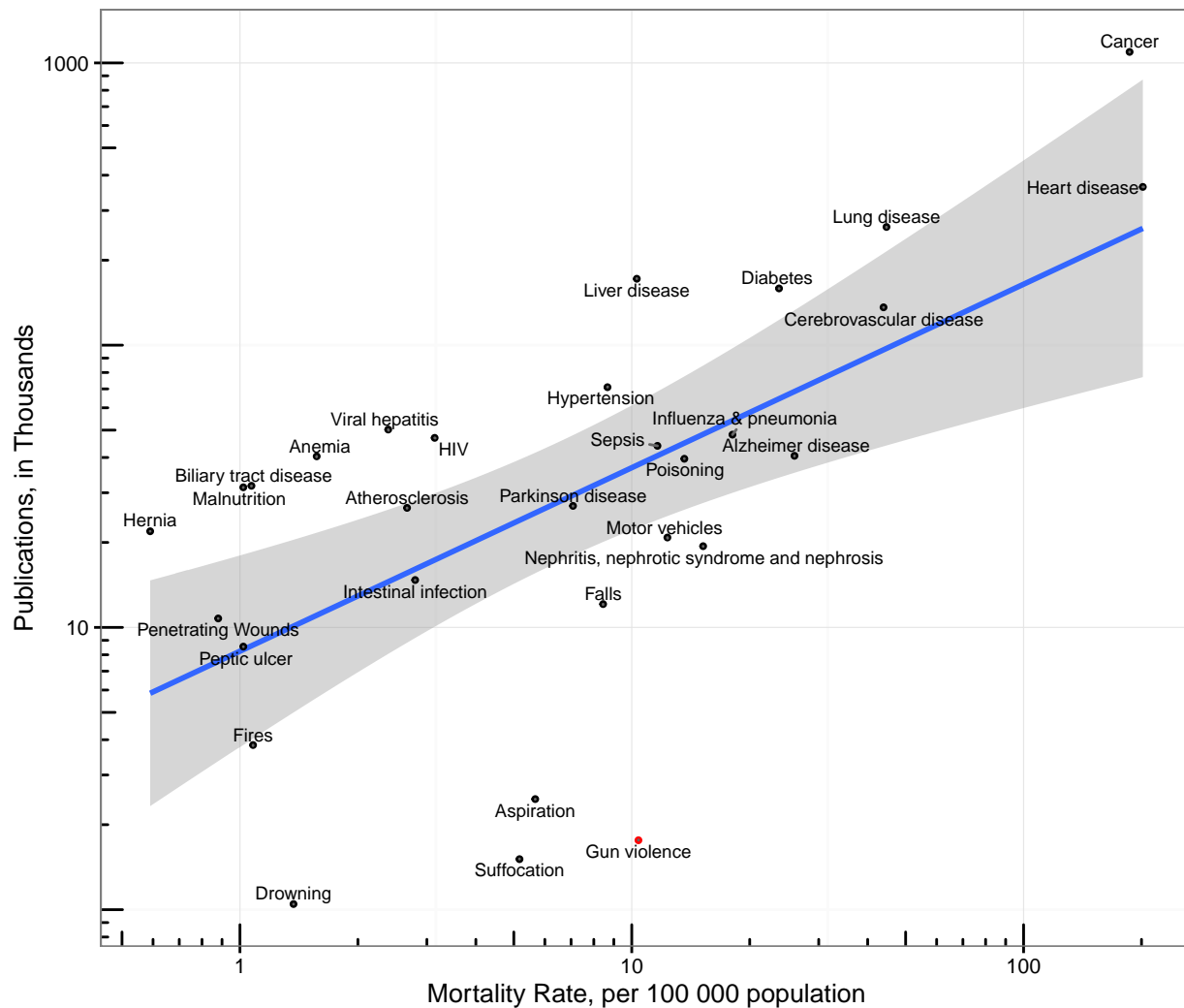


Figure 1. Mortality Rate vs Funding and Publication Volume for 30 Leading Causes of Death in the United States

HIV indicates human immunodeficiency virus. Shaded areas indicate 95% CIs. Plotting is on a log-log scale. Funding represents the total funding awarded over the years 2004 to 2015. Dollar amounts have not been correct for the year in which they were reported.

To determine how research funding and publication volume correlated with mortality, two linear regression analyses were performed using mortality rate as a predictor, and funding or publication count as outcomes. The predictor and outcomes were log-transformed and studentized residuals (residual divided by estimated standard error) were calculated to determine the extent to which a given cause of death was an outlier in terms of research funding or publication volume.

Code to calculate predicted funding, publication volume, and studentized residuals

```
# Regress mortality rate on publications, calculate predicted values and  
# residuals
```

```
lm.fit <- lm(log(Publications) ~ log(Crude.Rate), data = data)  
data$Publications.Predicted <- predict(lm.fit)
```



```

data$Publications.Residuals <- rstudent(lm.fit)
summary(lm.fit)

##
## Call:
## lm(formula = log(Publications) ~ log(Crude.Rate), data = data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.0644 -0.6602  0.2092  0.9952  1.5222
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    9.0163     0.3813  23.645 < 2e-16 ***
## log(Crude.Rate)  0.6501     0.1582   4.108 0.000314 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.308 on 28 degrees of freedom
## Multiple R-squared:  0.3761, Adjusted R-squared:  0.3538
## F-statistic: 16.88 on 1 and 28 DF,  p-value: 0.0003142
# Regress mortality rate on funding, calculate predicted values and
# residuals

lm.fit <- lm(log(Total.Funding) ~ log(Crude.Rate), data = data)
data$Funding.Predicted <- predict(lm.fit)
data$Funding.Residuals <- rstudent(lm.fit)
summary(lm.fit)

##
## Call:
## lm(formula = log(Total.Funding) ~ log(Crude.Rate), data = data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -5.9742 -0.9966  0.6406  1.1267  3.5047
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    19.5000     0.5737  33.988 < 2e-16 ***
## log(Crude.Rate)  0.7194     0.2381   3.022 0.00533 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.968 on 28 degrees of freedom
## Multiple R-squared:  0.2459, Adjusted R-squared:  0.219
## F-statistic:  9.13 on 1 and 28 DF,  p-value: 0.005325

```

Table 1: Publication Residuals.

Cause of Death	Publications	Predicted Publications	Residual
Heart disease	363,597	259,276	0.29
Cancer	1,093,740	246,484	1.29
Lung disease	262,290	97,373	0.79

Cause of Death	Publications	Predicted Publications	Residual
Cerebrovascular disease	136,163	96,279	0.27
Alzheimer disease	40,555	68,516	-0.41
Diabetes	158,876	64,586	0.7
Influenza & pneumonia	48,183	54,057	-0.09
Nephritis, nephrotic syndrome and nephrosis	19,400	48,329	-0.71
Poisoning	39,578	44,982	-0.1
Motor vehicles	20,825	42,164	-0.54
Sepsis	43,981	40,592	0.06
Gun violence	1,762	37,747	-2.63
Liver disease	171,877	37,510	1.19
Hypertension	70,967	33,536	0.58
Falls	12,097	32,980	-0.77
Parkinson disease	26,952	29,398	-0.07
Aspiration	2,465	25,446	-1.9
Suffocation	1,510	23,963	-2.31
HIV	46,946	17,329	0.77
Intestinal infection	14,722	16,085	-0.07
Atherosclerosis	26,517	15,595	0.41
Viral hepatitis	50,240	14,511	0.97
Anemia	40,379	11,042	1.03
Drowning	1,047	10,106	-1.88
Fires	3,831	8,658	-0.64
Biliary tract disease	31,753	8,606	1.04
Malnutrition	31,360	8,343	1.06
Peptic ulcer	8,552	8,343	0.02
Penetrating Wounds	10,759	7,579	0.28
Hernia	21,908	5,844	1.08

Table 2: Funding Residuals.

Cause of Death	Funding	Predicted Funding	Residual
Heart disease	\$ 13,923,184,996	\$ 13,381,192,791	0.02
Cancer	\$ 24,430,350,065	\$ 12,652,597,908	0.37
Lung disease	\$ 12,642,105,350	\$ 4,527,146,599	0.54
Cerebrovascular disease	\$ 8,337,570,924	\$ 4,470,882,290	0.33
Alzheimer disease	\$ 6,483,893,333	\$ 3,068,341,602	0.39
Diabetes	\$ 16,111,776,963	\$ 2,874,188,953	0.9
Influenza & pneumonia	\$ 1,604,400,094	\$ 2,360,403,894	-0.2
Nephritis, nephrotic syndrome and nephrosis	\$ 641,456,983	\$ 2,085,229,711	-0.61
Poisoning	\$ 657,603,508	\$ 1,926,002,522	-0.55
Motor vehicles	\$ 433,792,460	\$ 1,792,950,917	-0.73
Sepsis	\$ 3,502,000,967	\$ 1,719,124,966	0.36
Gun violence	\$ 23,506,911	\$ 1,586,291,049	-2.35
Liver disease	\$ 5,344,800,115	\$ 1,575,303,224	0.63
Hypertension	\$ 5,207,302,530	\$ 1,391,680,435	0.68
Falls	\$ 3,474,852	\$ 1,366,184,104	-3.74
Parkinson disease	\$ 3,468,984,129	\$ 1,202,934,400	0.54
Aspiration	\$ 2,766,427,514	\$ 1,025,309,670	0.51

Cause of Death	Funding	Predicted Funding	Residual
Suffocation	\$ 139,874,715	\$ 959,427,672	-1
HIV	\$ 22,299,391,060	\$ 670,219,322	1.9
Intestinal infection	\$ 512,893,132	\$ 617,178,692	-0.09
Atherosclerosis	\$ 4,610,352,415	\$ 596,427,288	1.07
Viral hepatitis	\$ 1,738,054,577	\$ 550,737,242	0.59
Anemia	\$ 2,388,220,073	\$ 407,055,883	0.93
Drowning	\$ 24,209,537	\$ 369,045,762	-1.46
Fires	\$ 773,841,193	\$ 311,005,331	0.48
Biliary tract disease	\$ 393,279,229	\$ 308,930,970	0.13
Malnutrition	\$ 2,060,162,078	\$ 298,476,119	1.03
Peptic ulcer	\$ 145,347,374	\$ 298,476,119	-0.38
Penetrating Wounds	\$ 87,951,989	\$ 268,400,332	-0.59
Hernia	\$ 93,904,298	\$ 201,313,101	-0.41

Code for residuals plot

```
# plot funding residuals x publication residuals
fig2 <- ggplot(data, aes(x = Publications.Residuals, y = Funding.Residuals)) +
  geom_hline(yintercept = 0, size = 0.5, color = "gray") + geom_vline(xintercept = 0,
  size = 0.5, color = "gray") + geom_point(size = 0.75, color = as.numeric(data$Abbreviation ==
  "Gun violence") + 1) + scale_color_brewer(type = "qual", palette = "Set1") +
  geom_text_repel(aes(label = Abbreviation), size = 2.4, segment.size = 0.5,
    box.padding = unit(0.1, "lines")) + theme_bw(base_size = 10) + labs(y = "Funding, Studentized Residuals",
  labs(x = "Publications, Studentized Residuals")
fig2
```

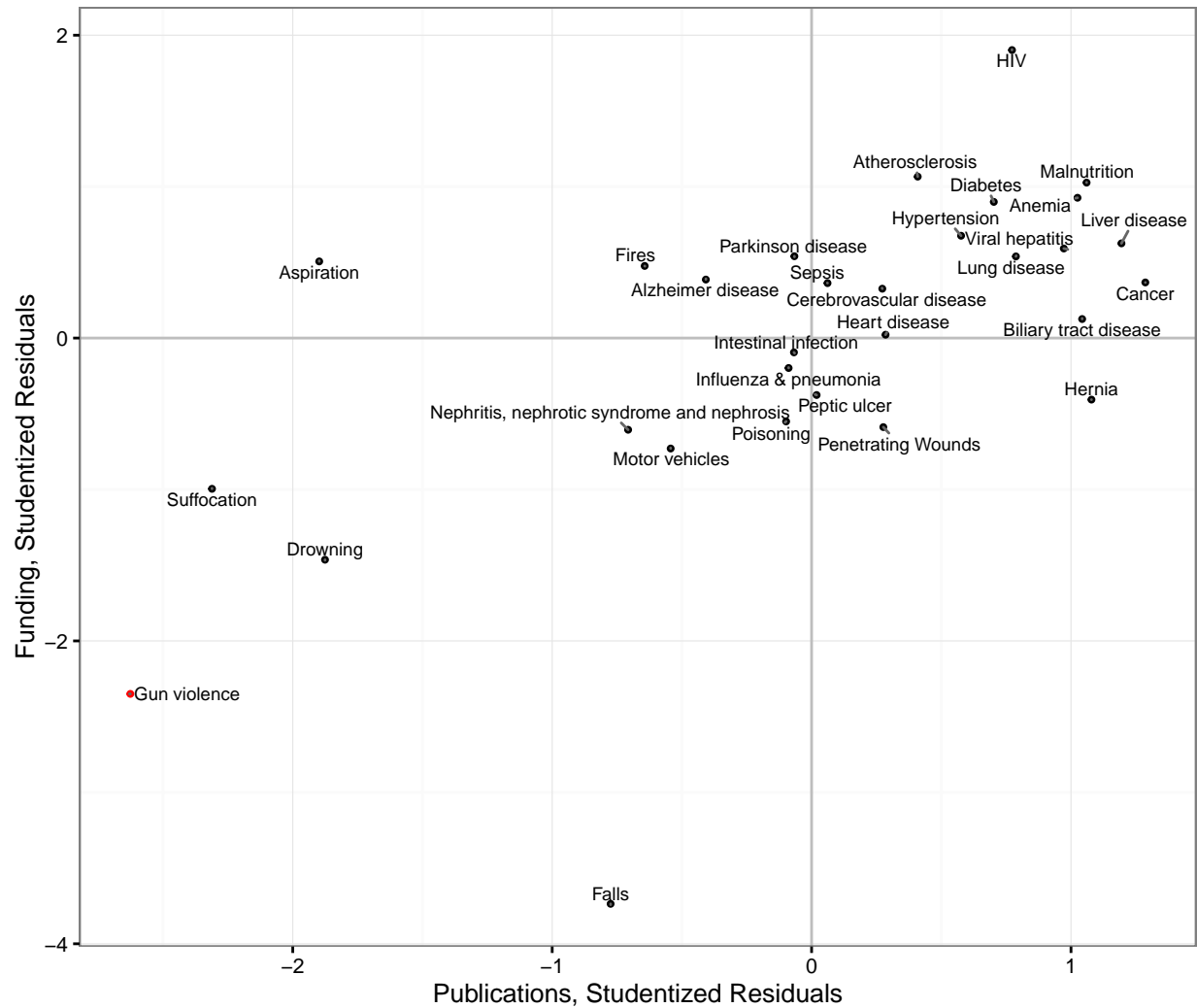


Figure 2. Studentized Residual Predicted vs Observed Funding and Publication Volumes for 30 Leading Causes of Death in the United States

HIV indicates human immunodeficiency virus. Mortality rate was used to predict funding and research volume. Studentized residuals (residual divided by estimated standard error) were calculated to give a standardized estimate of predicted vs observed funding and publication volume. The 4 quadrants represent observed funding greater than predicted, observed publication volume less than predicted (upper-left); observed funding and publication volume greater than predicted (upper-right); observed funding less than predicted, observed publication volume greater than predicted (lower-right); observed funding and publication volume less than predicted (lower-left).

““