

# Do More People Die of Extreme Heat or Excessive Cold?

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**Synopsis:** The question of “Do More People Die of Extreme Heat or Excessive Cold?” is addressed using the “**R**” statistical language and data from the **National Oceanic and Atmospheric Administration (NOAA)** “**Storm Data**” database.

**Data Processing** The **R** language is an open source version of the the **S** language developed at Bell Labs during the “golden age” that also produced the **Unix** operating system and the **C** programming language. See: <https://www.R-project.org/>

Unlike **C**, **R** is an interpretive command language where the user types commands at the command line and gets an immediate response:

```
2+2
```

```
## [1] 4
```

```
sqrt(25)
```

```
## [1] 5
```

```
GaussDidThisInHisHeadInElemetarySchool <- sum(1:100)
print(GaussDidThisInHisHeadInElemetarySchool)
```

```
## [1] 5050
```

The last of these three examples is a problem solved by famous mathematician **Carl Friedrich Gauss (1777-1855)** while he was an eight year old child math prodigy in elementary school. He solved it in his head, amazing his teacher. For more of the story see: “Clever Carl” <http://nrich.maths.org/2478/index?nomenu=1>

For those of us who are not (child or adult) math prodigies we can solve the problem with **R** by typing **sum(1:100)** at the command line. The “<-” assigns the result of the function to the variable name on the left.

The command line commands can be combined in simple text files “scripts” or combined with compiled programs, data and documentation to form complete “packages”.

Many open source statistical “packages” have been written in **R** making the complete system, the base language plus the optional downloadable statistical packages more powerful than proprietary statistical systems such as SAS or SPSS which, depending on the license, whose base price can be as much as \$9,000 a seat. See: <http://www.sas.com/store/products-solutions/cSoftware-pl.html>

When we load the **NOAA “Storm Data” data file** into the **R** statistical system, we will also be using the “<-” to assign the result to the variable name on the left.

```
filename <- "~/GitHub/RepData_PeerAssessment2/data/repdata%2Fdata%2FStormData.csv.bz2"
NOAA <- read.csv(filename,
                  stringsAsFactors = FALSE )
```

```

# Select Columns of interest:
#   Primary key:      REFNUM
#   Date and Time:    BGN_DATE, BGN_TIME, TIME_ZONE,
#   Location:         STATE, COUNTY, COUNTY_END, LATITUDE, LONGITUDE,
#   Type of Storm:    EVTYPE,
#   Damage:           PROPDMG, PROPDMGEXP, CROPDMG, CROPDMGEXP,
#   Casualties:       FATALITIES, INJURIES

columnsubset <- c("REFNUM", "BGN_DATE", "BGN_TIME", "TIME_ZONE",
                  "STATE", "COUNTY", "COUNTY_END", "LATITUDE", "LONGITUDE",
                  "EVTYPE", "PROPDMG", "PROPDMGEXP", "CROPDMG", "CROPDMGEXP",
                  "FATALITIES", "INJURIES")

storms <- NOAA[, columnsubset]
str(storms)

```

```

## 'data.frame':   902297 obs. of  16 variables:
## $ REFNUM      : num  1 2 3 4 5 6 7 8 9 10 ...
## $ BGN_DATE    : chr  "4/18/1950 0:00:00" "4/18/1950 0:00:00" "2/20/1951 0:00:00" "6/8/1951 0:00:00" ...
## $ BGN_TIME    : chr  "0130" "0145" "1600" "0900" ...
## $ TIME_ZONE   : chr  "CST" "CST" "CST" "CST" ...
## $ STATE       : chr  "AL" "AL" "AL" "AL" ...
## $ COUNTY      : num  97 3 57 89 43 77 9 123 125 57 ...
## $ COUNTY_END  : num  0 0 0 0 0 0 0 0 0 0 ...
## $ LATITUDE    : num  3040 3042 3340 3458 3412 ...
## $ LONGITUDE   : num  8812 8755 8742 8626 8642 ...
## $ EVTYPE      : chr  "TORNADO" "TORNADO" "TORNADO" "TORNADO" ...
## $ PROPDMG     : num  25 2.5 25 2.5 2.5 2.5 2.5 2.5 25 25 ...
## $ PROPDMGEXP  : chr  "K" "K" "K" "K" ...
## $ CROPDMG     : num  0 0 0 0 0 0 0 0 0 0 ...
## $ CROPDMGEXP  : chr  "" "" "" "" ...
## $ FATALITIES  : num  0 0 0 0 0 0 0 0 1 0 ...
## $ INJURIES    : num  15 0 2 2 2 6 1 0 14 0 ...

```

```

#### Select Rows of interest
#### Select years of interest (Since 2001 "21st Century Storms" or "last 25 years")

#### Step 1: Create a year variable from date
#### Read in the date.
#### NOAA's Time and time zone are in separate variables that we will ignore.
datetime = as.POSIXct(storms$BGN_DATE, "%m/%d/%Y %H:%M:%S", tz = "")
storms$year <- format(datetime, "%Y")

#### Step 2: Filter by year
storms <- storms[storms$year >= "2001", ]

#### Remove the original NOAA database
rm(NOAA)

```

```

#### Scale Property Damage and Crop Damage by thousands and millions
storms$PropertyDamage <- storms$PROPDMG
storms$CropDamage      <- storms$CROPDMG

```

```

with(storms,
{
  PropertyDamage <- ifelse(PROPDMGEXP == "K", PROPDMG * 1000, PROPDMG)
  PropertyDamage <- ifelse(PROPDMGEXP == "M", PROPDMG * 1000 * 1000, PROPDMG)
  # Round off total to nearest dollar.
  PropertyDamage <- round(PropertyDamage, digits = 0)

  CropDamage <- ifelse(CROPDMGEXP == "K", CROPDMG * 1000, CROPDMG)
  CropDamage <- ifelse(CROPDMGEXP == "M", CROPDMG * 1000 * 1000, CROPDMG)
  # Round off total to nearest dollar.
  CropDamage <- round(CropDamage, digits = 0)
}
)

#### Rename and put variables in logical order.

```

## Results

```

#### Tabulate Fatalities and Damage by Storm Type and Year
#### Useful R commands include table(), xtabs(), ftable() or aggregate()
#### This use of aggregate() is based on Jared Lander's "R for Everyone" page 123
#### where he uses aggregate() on the diamonds data set from the ggplot2 package.
stormstab <- aggregate(cbind(FATALITIES, INJURIES, PropertyDamage, CropDamage)
~ EVTYPE + year, storms, sum)

```

## Tabulate Fatalities and Damage by Storm Type and Year

```

# What type of storms caused the most fatalities in the most recent year (2011)?
mostrecentyear <- max(stormstab$year)
stormsrankyear <- stormstab[stormstab$year == mostrecentyear, ]

# This is the sort -- rank by fatalities in the stormrankyear
deadlystorms <- stormsrankyear[order(-stormsrankyear$FATALITIES, na.last = NA), ]
# head(deadlystorms, 10)
print(deadlystorms[1:10, c("EVTYPE", "FATALITIES", "INJURIES", "PropertyDamage", "CropDamage")],
      justify = "left")

```

## Health Impact

##	EVTYPE	FATALITIES	INJURIES	PropertyDamage	CropDamage
## 630	TORNADO	587	6163	155464.51	18374
## 603	FLASH FLOOD	68	30	82067.98	5530
## 609	HEAT	63	611	0.00	0
## 604	FLOOD	58	10	133748.97	25002
## 629	THUNDERSTORM WIND	54	373	243100.34	16955
## 601	EXCESSIVE HEAT	36	138	1143.20	0

## 624	RIP CURRENT	29	27	0.00	0
## 619	LIGHTNING	26	194	34491.42	112
## 595	COLD/WIND CHILL	21	1	70.00	0
## 612	HIGH SURF	11	11	222.00	0

```
# What type of storms caused the most fatalities in the most recent year (2011)?
damagestorms <- stormsrankyear[order(-stormsrankyear$PropertyDamage, na.last = NA), ]
head(damagestorms, 10)
```

## Property Damage

##	EVTYPE	year	FATALITIES	INJURIES	PropertyDamage	CropDamage
## 629	THUNDERSTORM WIND	2011	54	373	243100.34	16955
## 630	TORNADO	2011	587	6163	155464.51	18374
## 604	FLOOD	2011	58	10	133748.97	25002
## 603	FLASH FLOOD	2011	68	30	82067.98	5530
## 608	HAIL	2011	0	31	57214.06	26390
## 619	LIGHTNING	2011	26	194	34491.42	112
## 634	WILDFIRE	2011	6	116	33094.24	1805
## 613	HIGH WIND	2011	4	11	14728.25	1336
## 631	TROPICAL STORM	2011	4	1	11769.30	525
## 635	WINTER STORM	2011	1	0	11663.50	70

Event Descriptions indicating Hot or Cold Weather

COLD  
 COLD/WIND CHILL  
 COLD WIND CHILL TEMPERATURES  
 EXCESSIVE HEAT  
 EXTREME COLD  
 EXTREME COLD/WIND CHILL  
 EXTREME WINDCHILL  
 FREEZE  
 FREEZING DRIZZLE  
 FREEZING RAIN  
 FROST/FREEZE  
 HARD FREEZE  
 HEAT  
 RECORD WARMTH  
 UNSEASONABLY COLD  
 UNSEASONABLY COOL  
 UNSEASONABLY WARM  
 UNUSUALLY COLD  
 WINTER STORM  
 WINTER WEATHER  
 WINTER WEATHER/MIX

END