Climate-Related Hazards Damage SHELDUS data

Data Source:

SPATIAL HAZARD EVENTS AND LOSSES DATABASE FOR THE UNITED STATES (SHELDUS) Version 23.0

https://sheldus.asu.edu/

Downloaded: July 3, 2025

Key Variables:

- cropding: Direct damage to crops in nominal U.S. dollars.
- cropdmg_adj: Direct damage to crops in adjusted 2023 U.S. dollars.
- CropDmgPerCapita: Crop damage per capita, adjusted to the most recent SHELDUS inflation base year.
- propertydmg: Direct damage to property in nominal U.S. dollars (no decimals).
- propertydmg_adj: Direct damage to property in adjusted 2023 U.S. dollars (two decimal places).
- PropertyDmgPerCapita: Property damage per capita, adjusted to the most recent SHEL-DUS inflation base year.
- duration_days: Duration of the event in days (integer).

Hazard Categories (18 total):

- 1. Avalanche
- 2. Coastal
- 3. Drought
- 4. Earthquake
- 5. Flood
- 6. Fog
- 7. Hail
- 8. Heat
- 9. Hurricane/Tropical Storm
- 10. Landslide
- 11. Lightning
- 12. Severe Thunderstorm
- 13. Tornado
- 14. Tsunami/Seiche
- 15. Volcano
- 16. Wildfire
- 17. Wind
- 18. Winter Weather Load Libraries and Data

```
library(tidyverse)
library(gt)
library(scales)

# Load and clean SHELDUS data
sheldus <- read_csv(here("data", "direct_loss_aggregated_output_5008.csv"))
names(sheldus) <- tolower(names(sheldus))

sheldus_clean <- sheldus %>%
    rename(
    state = statename,
        cropdmg_adj = `cropdmg(adj 2023)`,
        propertydmg_adj = `propertydmg(adj 2023)`
) %>%
    select(state, year, hazard, cropdmg = cropdmg_adj, propertydmg = propertydmg_adj)
```

Summary Statistics

Annual Damage Summary

```
summary_by_year <- sheldus_clean %>%
 group_by(year) %>%
 summarize(
   total_property_dmg = sum(propertydmg, na.rm = TRUE),
   total_crop_dmg = sum(cropdmg, na.rm = TRUE),
   .groups = "drop"
 )
summary_by_year |>
 mutate(
   total_property_dmg = total_property_dmg / 1e6,
   total_crop_dmg = total_crop_dmg / 1e6
 ) |>
 gt() |>
 tab header(title = "Annual Direct Damage Summary (Adjusted to 2023 Dollars)") |>
 fmt_currency(columns = starts_with("total"), currency = "USD") |>
 cols_label(
   total_property_dmg = "Total Property Damage (million)",
   total_crop_dmg = "Total Crop Damage (million)"
 )
```

Damage by State

```
summary_by_state <- sheldus_clean %>%
  group_by(state) %>%
  summarize(
    total_property_dmg = sum(propertydmg, na.rm = TRUE),
    total_crop_dmg = sum(cropdmg, na.rm = TRUE),
    .groups = "drop"
  ) %>%
  arrange(desc(total_property_dmg))

summary_by_state |>
  mutate(
    total_property_dmg = total_property_dmg / 1e6,
```

Annual Direct Damage Summary (Adjusted to 2023 Dollars)

year	Total Property Damage (million)	Total Crop Damage (million)
2000	\$9,919.50	\$5,817.19
2001	\$22,510.80	\$2,755.44
2002	\$6,781.59	\$2,094.61
2003	\$17,354.38	\$1,866.75
2004	\$40,425.09	\$2,189.22
2005	\$149,046.27	\$6,164.08
2006	\$10,656.68	\$5,124.74
2007	\$10,689.08	\$7,090.95
2008	\$37,893.56	\$4,813.12
2009	\$8,093.02	\$934.26
2010	\$13,921.90	\$3,384.90
2011	\$31,204.97	\$4,405.15
2012	\$43,014.49	\$8,070.91
2013	\$11,876.21	\$5,661.41
2014	\$7,998.84	\$4,670.56
2015	\$5,606.35	\$800.62
2016	\$22,632.70	\$299.36
2017	\$100,313.59	\$2,309.18
2018	\$39,106.44	\$6,158.66
2019	\$19,529.16	\$1,632.49
2020	\$31,023.41	\$583.50
2021	\$19,921.82	\$915.90
2022	\$22,284.20	\$1,085.60
2023	\$16,631.10	\$1,092.73

```
total_crop_dmg = total_crop_dmg / 1e6
) |>
gt() |>
tab_header(title = "Total Damage by State (Adjusted to 2023 Dollars)") |>
fmt_currency(columns = starts_with("total"), currency = "USD") |>
cols_label(
   total_property_dmg = "Total Property Damage (million)",
   total_crop_dmg = "Total Crop Damage (million)"
)
```

Damage by Hazard Type

```
summary_by_hazard <- sheldus_clean %>%
 group_by(hazard) %>%
 summarize(
   total_property_dmg = sum(propertydmg, na.rm = TRUE),
   total_crop_dmg = sum(cropdmg, na.rm = TRUE),
    .groups = "drop"
 ) %>%
 arrange(desc(total_property_dmg))
summary_by_hazard |>
 mutate(
   total_property_dmg = total_property_dmg / 1e6,
   total_crop_dmg = total_crop_dmg / 1e6
 ) |>
 gt() |>
 tab header(title = "Total Damage by Hazard Type (Adjusted to 2023 Dollars)") |>
 fmt_currency(columns = starts_with("total"), currency = "USD") |>
 cols_label(
   total_property_dmg = "Total Property Damage (million)",
   total_crop_dmg = "Total Crop Damage (million)"
 )
```

Climate-Related Hazard Focus

Damage Trends: Most Climate-Related Hazards, including "Drought", "Flood", "Heat", "Hurricane/Tropical Storm", "Wildfire", "Coastal"

Total Damage by State (Adjusted to 2023 Dollars)

Total Property Damage (million)

\$132,000.92

\$130,601.72

\$85,349.94

state

TEXAS

LOUISIANA

FLORIDA

CALIFORNIA	\$52,999.60
MISSISSIPPI	\$45,057.24
NEW JERSEY	\$35,611.89
PUERTO RICO	\$26,343.10
ALABAMA	\$14,931.64
TENNESSEE	\$11,000.95
MISSOURI	\$10,994.86
COLORADO	\$9,960.99
OKLAHOMA	\$9,373.65
WASHINGTON	\$9,161.63
OHIO	\$8,633.29
NORTH CAROLINA	\$8,242.54
MICHIGAN	\$6,849.81
ARKANSAS	\$6,381.35
ILLINOIS	\$6,143.43
NEW YORK	\$6,130.04
HAWAII	\$6,091.89
GEORGIA	\$5,609.90
IOWA	\$5,485.61
ARIZONA	\$5,443.74
OREGON	\$4,921.02
GUAM	\$4,250.22
NEBRASKA	\$4,230.67
PENNSYLVANIA	\$3,966.31
NEW MEXICO	\$3,898.89
WISCONSIN	\$3,872.33
KANSAS	\$3,812.52
MINNESOTA	\$3,432.67
KENTUCKY	\$2,939.95
INDIANA	\$2,866.21
VIRGINIA	\$2,710.18
VERMONT	\$2,160.55
NORTH DAKOTA	\$2,004.93
MARYLAND	\$1,827.91
WEST VIRGINIA	\$1,564.19
DISTRICT OF COLUMBIA	\$1,505.17
UTAH	\$1,384.29
SOUTH CAROLINA 6	\$1,312.36
IDAHO	\$1,138.97
MASSACHUSETTS	\$1,122.63
SOUTH DAKOTA	\$880.26
ALASKA	\$781.13
COMMONWEALTH OF THE NORTHERN MARIANA ISLANDS	\$645.60
NEVADA	\$503.96

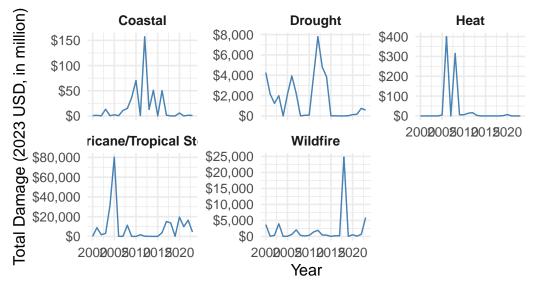
Total Damage by Hazard Type (Adjusted to 2023 Dollars)

hazard	Total Property Damage (million)	Total Crop Damage (million)
Flooding	\$278,109.33	\$11,921.20
Hurricane/Tropical Storm	\$208,123.52	\$12,746.90
Wildfire	\$47,771.32	\$543.11
Tornado	\$46,648.56	\$462.14
Hail	\$38,764.72	\$4,149.43
Wind	\$27,403.43	\$2,540.72
Earthquake	\$20,578.82	\$0.00
Winter Weather	\$15,434.16	\$8,274.82
Severe Storm/Thunder Storm	\$6,541.55	\$1,642.13
Landslide	\$3,845.13	\$29.92
Drought	\$3,298.01	\$36,837.79
Lightning	\$1,243.52	\$7.19
Coastal	\$432.07	\$1.22
Tsunami/Seiche	\$103.82	\$0.00
Fog	\$62.98	\$0.00
Heat	\$41.35	\$730.90
Volcano	\$19.36	\$33.89
Avalanche	\$13.48	\$0.00

```
most_climate_related_hazards <- c(</pre>
  "Drought",
  "Flood",
  "Heat",
  "Hurricane/Tropical Storm",
  "Wildfire",
  "Coastal"
)
plot_top_hazards <- sheldus_clean %>%
  filter(hazard %in% most_climate_related_hazards) %>%
  mutate(total_dmg = propertydmg + cropdmg) %>%
  mutate(total_dmg = total_dmg / 1e6) %>%
  group_by(year, hazard) %>%
  summarize(total_dmg = sum(total_dmg, na.rm = TRUE), .groups = "drop") %>%
  ggplot(aes(x = year, y = total_dmg)) +
  geom_line(color = "steelblue") +
  facet_wrap(~ hazard, scales = "free_y") +
  scale_y_continuous(labels = label_dollar()) +
  labs(
    title = "Most Climate-Related Hazards: Annual Damage Trends",
    subtitle = "Total direct damage (property + crop), adjusted to 2023 dollars",
   x = "Year",
    y = "Total Damage (2023 USD, in million)"
  theme_minimal(base_size = 12) +
    strip.text = element_text(face = "bold"),
    plot.title = element_text(face = "bold")
  )
plot_top_hazards
```

Most Climate-Related Hazards: Annual Damage Tr

Total direct damage (property + crop), adjusted to 2023 dollars



Data ready for merge

```
sheldus_formerge <- sheldus |>
 select(
   statename,
    year,
   hazard,
    cropdmg,
    "cropdmg(adj 2023)",
    cropdmgpercapita,
    propertydmg,
    "propertydmg(adj 2023)",
   propertydmgpercapita,
    duration_days
 ) |>
 rename(
    cropdmg_adj = "cropdmg(adj 2023)",
    propertydmg_adj = "propertydmg(adj 2023)",
    state = statename
 )
```

summary(sheldus_formerge)

state	year	hazard	cropdmg	
Length:9563	Min. :2000 Le	ngth:9563	Min. :0.000e+00	
Class :character	1st Qu.:2006 Cl	ass :character	1st Qu.:0.000e+00	
Mode :character	Median:2011 Mo	de :character	Median :0.000e+00	
	Mean :2011		Mean :6.048e+06	
	3rd Qu.:2017		3rd Qu.:3.000e+03	
	Max. :2023		Max. :3.000e+09	
cropdmg_adj	cropdmgpercapita	. propertydmg	ξ	
Min. :0.000e+00	Min. : 0.0	0 Min. :0.00)0e+00	
1st Qu.:0.000e+00	1st Qu.: 0.0	0 1st Qu.:3.10	00e+04	
Median :0.000e+00	Median: 0.0	0 Median :5.59	92e+05	
Mean :8.357e+06	Mean : 649.4	0 Mean :5.45	59e+07	
3rd Qu.:3.696e+03	3rd Qu.: 0.0	6 3rd Qu.:3.53	37e+06	
Max. :3.815e+09	Max. :525557.6	4 Max. :4.41	1e+10	
	NA's :35			
<pre>propertydmg_adj</pre>	propertydmgperca	pita duration_da	ıys	
Min. :0.000e+00	Min. : 0.0	0 Min. : 1	.000	
1st Qu.:4.210e+04	1st Qu.: 0.5	4 1st Qu.: 1	.000	
Median :7.510e+05	Median: 16.2	6 Median: 1	.000	
Mean :7.304e+07	Mean : 1240.4	1 Mean : 2	2.985	
3rd Qu.:4.942e+06	3rd Qu.: 133.1	0 3rd Qu.: 2	2.000	
Max. :5.324e+10	Max. :829879.7	0 Max. :351	.000	
	NA's :35			

Descriptive Statistics

```
library(psych)
describe(sheldus_formerge |> select(-state, -year, -hazard))
```

	vars	n	mean	sd	median	trimmed
cropdmg	1	9563	6048100.34	7.905309e+07	0.00	17319.04
cropdmg_adj	2	9563	8357352.75	1.062094e+08	0.00	23703.55
${\tt cropdmgpercapita}$	3	9528	649.40	1.038799e+04	0.00	0.95
propertydmg	4	9563	54585404.69	8.308901e+08	559250.00	2172119.32
propertydmg adi	5	9563	73035148.93	1.110126e+09	751000.00	2960802.99

propertydmgpercapita	6 9528		1240.41	1.729	9825e+04	16.26	83.46
duration_days	7 9563		2.99	7.340	0000e+00	1.00	1.54
	mad m	in		max	range	skew	kurtosis
cropdmg	0.0	0	3.00000	0e+09	3.000000e+09	24.92	753.15
cropdmg_adj	0.0	0	3.81513	7e+09	3.815137e+09	23.52	667.98
${\tt cropdmgpercapita}$	0.0	0	5.25557	6e+05	5.255576e+05	31.41	1194.07
propertydmg	829144.1	0	4.410998	Be+10	4.410998e+10	31.73	1265.27
<pre>propertydmg_adj</pre>	1113432.6	0	5.32397	7e+10	5.323977e+10	31.13	1177.35
${\tt propertydmgpercapita}$	24.1	0	8.29879	7e+05	8.298797e+05	33.55	1336.44
duration_days	0.0	1	3.510000	0e+02	3.500000e+02	23.85	1058.89
	se						
cropdmg	808391.60						
cropdmg_adj	1086090.26						
${ t cropdmgpercapita}$	106.42						
propertydmg	8496625.90						
<pre>propertydmg_adj</pre>	11352075.88						
${\tt propertydmgpercapita}$	177.22						
duration_days	0.08						