

Econ/Demog C175 Lab 4: The Bet – Over and over again

Code ▾

Julian Simon liked to proclaim that

The only true measure of scarcity is price ...

In this lab we are going to take advantage of a database put together by David Jacks. The data for his paper “From Boom to Bust: A Typology of Real Commodity Prices in the Long Run” (2013, NBER Working Paper 18874) has consistent series of commodity prices for everything ranging from barley to zinc. For our lab we’ve selected prices since 1900.

We’re going to ask the following:

1. What has been the long-term trend in commodity prices?
2. Was Simon lucky to have won the bet?
3. Do commodity price trends differ by commodity in a way that we can understand? (This section will be more exploratory than in past labs. Here the important point is to report *what* you find, not to find a specific “right” answer.)
4. Finally, we will use our Malthusian “harvest” model to learn about renewable resources. (We may or may not get to this in class this week. If not, then this part of the lab will be pushed into next week, when we have no lab).

Hide

```
# Do not edit this chunk, but *do* press the green button to the answer key for the quiz
info (the unreadable string below)
tot = 0
answer.key = "eJzNVk2P4zYMvedXEDmMd4FASZzMYHeBwSiZLdBeiQKz94EiM7YaWfLqYz359yVlJ/Phaw69GK
Fok3yP5FPET5vaPfq1WMM90HMmlPMEvRTShv7smP82n4lG2ziaf3utMID0CNpW+IIVRAfrr6sVSFv1c1n9m0Ixx8
F5eulgZNTOUhhTQ/RJsRXGcP9gZ6RCiA3CXzvodWzg5JKHoQb4hKIWCyh2BUWCn8lFDJ8pFL7QdzYHvkSSBrqxul
zBJP8FcZkRl1PE7Jg/XBBns4mxC9+WS7Si10fdYaWlcL5esrV8bLxrdWon8MqrwcuRdopqrbStme5LHQv4cgsdeo
U2gibgXUcdsNGcIAWsZoIpiA15nEXoku9cQAE7n0uiz0j7EHVmlDfT1XkqgSBQHmohmu+vtJWZ53I6KINj/jjSNp
pP9M5pAdYBH36gp7xa98dIPxhNEUAZlJ7QSzAugjvAL8ftNzqeCLaJjUt1k/PlUeGw5Wp9y9wluqYg5JOWj3mkFw
N3lcNgi0gh3RGMPTkID4kgEtnVAoK2Fiv5a5NqOC1nKFcEhzuT/OlNy14GitR/SCXeElmoifzOTjmF56z9WdtHS
9diq4lpApaDEHwdLJ3Kb6jc4/c14KKHcMWM5GJY/jUqGI8L8YPBDykTKKiIIFR03gcKDwB7V0yFRAJdiai66WvoH
gsJm2+1hb8T8FeOrbJHdtM07Z5qyiJuTmmC6VHGZyVe4MLwBeFXyRH6BtNI0Q1v1YOP2iAZC2JHH1gkFDpigdyhP
ogJrxvrsb7UPFFV2g16N1RZ3umUzkbvTOGBJ890kfY8wIq+oWGGsUiq6TQBwFPIHAW1L7vhY5eE2ciHZf5OhFd03
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nhQEQofLN2m6wkm6m+DY5LE0fzj0EuRiZamgloetYD4VYSNx96srma5I2RdoxK8QxlGRSLCamuMcQwgL64Z6I4F1
YsaMzkhn8bb3jUWI/2GLPmO/uOkDITMtGhzXglXQgZdtMGpN6T8mqL36Cggmz85LyutZVGUApB5CcTP2doTIByFT
2aZI98qwUcaBneOqsofzeh8lqyckYCFms6+IUwQM/9DK9jTcrj94i/8cJDp50S8M1yEIj6bP+HXPb3KHtdJS270
dpNEfmhhuZhWG9umPcga/PSNbqIwPbqW3TG0muXHJS/guRlXQlbr+I+X+1gFbM"
library(quizify)
source.coded.txt(answer.key)
```

Note: the most helpful reading for this lab is David Lam (2011), “How the World Survived the Population Bomb”

Part 1. Read in data and graph trends in commodity prices

These commodity prices are modified in two ways. First, the nominal prices have been adjusted for inflation. Second, all prices are indexed to 100 in 1900. This makes it easy for us to compare across commodities and over time.

[Hide](#)

```
## get commodity data from a file on our server
file <- "/data175/Real_commodity_prices_1900_2015.csv"
commod <- read.table(file, sep = ",") # read the data into, with column header
head(commod) # looks at first few lines
```

	Beef <dbl>	Hides <dbl>	Lamb <dbl>	Pork <dbl>	Coal <dbl>	Natural.gas <dbl>	Petroleum <dbl>	Barley <dbl>	Corn <dbl>
1900	100.00	100.00	100.00	100.00	100.00	100.00	100.00	100.00	100.00
1901	98.79	95.25	90.41	115.35	99.74	79.03	88.37	110.19	129.32
1902	97.60	94.59	90.98	133.95	105.11	78.33	89.28	108.63	152.91
1903	90.87	88.12	93.81	113.47	113.78	76.66	112.12	104.81	114.99
1904	94.32	95.05	94.32	95.96	99.76	74.38	113.50	95.48	124.54
1905	95.10	119.72	95.43	98.74	97.26	66.80	97.69	92.38	125.36

6 rows | 1-10 of 40 columns

You can scroll to the right and left using the black arrows in the display of the data.

Q1.1 What does the number 110.51 under Zinc mean?

- A. Zinc costs 110.51 cents per pound
- B. Zinc prices increased by about 110 percent between 1900 and 1904
- C. Zinc cost about 10.51 percent more in *nominal* terms in 1904 than in 1900
- D. Zinc cost 10.51 percent more in *real* terms in 1904 than in 1900

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```
## "Replace the NA with your answer (e.g., 'A' in quotes)"
answer1.1 = "D"
quiz.check(answer1.1)
```

```
Your answer1.1 : D
Correct.
Explanation: Real prices adjust for inflation
```

Q1.2 What is the major use of Chromium?

- A. For browsing the internet
- B. As Chrome plating and stainless steel
- C. As vitamin supplement for breakfast cereal
- D. As a precious metal for storing value

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```
## "Replace the NA with your answer (e.g., 'A' in quotes)"
answer1.2 = "B"
quiz.check(answer1.2)
```

```
Your answer1.2 : B
Correct.
Explanation: According to wikipedia, 85 percent is apparently used
.for this one purpose. Are there substitutes for protecting steel?
```

To select a value from this matrix of data you can index by the label of the row and column. For example,

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```
commod["1963", "Chromium"]
```

```
[1] 55.07
```

Hide

```
## note for those interested in R:
## the index "1963" needs to be in quotes, otherwise R will think it
## is the 1963rd row of the matrix. Here we're telling it that it is
## the row labeled "1963".
```

tells us that the real price of Chromium in 1963 was 55 percent of that in 1900.

(Note: in the rest of the lab, when we say “price” we mean “real price”, unless otherwise stated.)

Part 2. Visualizing commodity prices

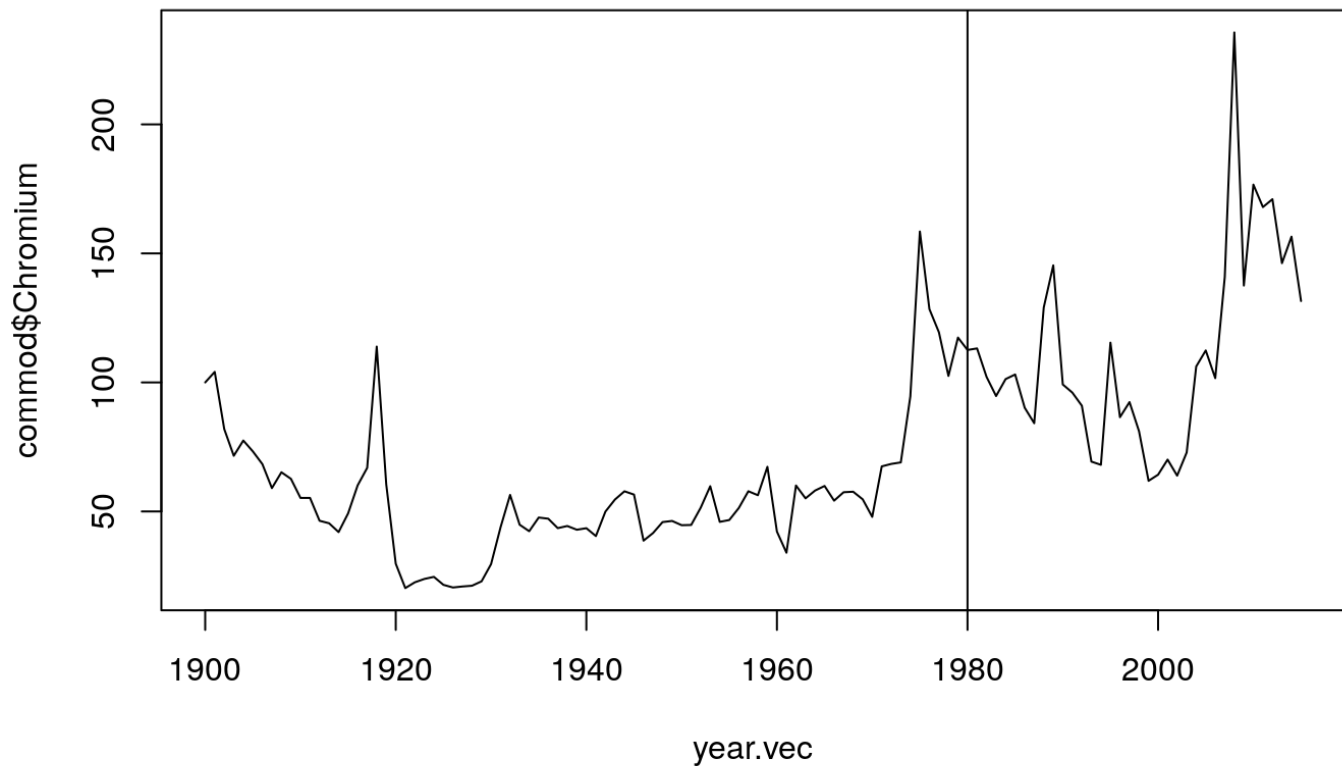
Let's see how the price of Chromium varies over time.

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```

year.vec <- rownames(commod)
plot(year.vec, commod$Chromium, type = "l")
abline(v = 1980)                                # the date of "the bet"

```



Q2.1 How would you describe the history of Chromium prices?

- A. Steadily rising
- B. Steadily falling
- C. Very volatile, with little clear long-term trend
- D. Very volatile, with a clear upward trend

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```

## "Replace the NA with your answer (e.g., 'A' in quotes)"
answer2.1 = "C"
quiz.check(answer2.1)

```

Your answer2.1 : C

Correct.

Explanation: There's clearly a lot of volatility. Although the price in 2015 is higher than in 1900, this doesn't look like a clear trend, since for much of the 20th century the price was lower than in 1900.

What does “trendless” data look like?

Imagine a “random walk”, in which prices go up or down each year by some random amount that averages zero. Each realization of this random walk will have, by luck, some trend, but the underlying process generating the randomness has no trend – on average the change is “zero”. It is very easy to falsely interpret a random walk as having an underlying trend. The subject of how to make inferences about the trend of a time series is covered in an econometrics or statistics class. For now, we will just look at some realizations of the random walk to get a feeling for what random realizations of trendless data can look like.

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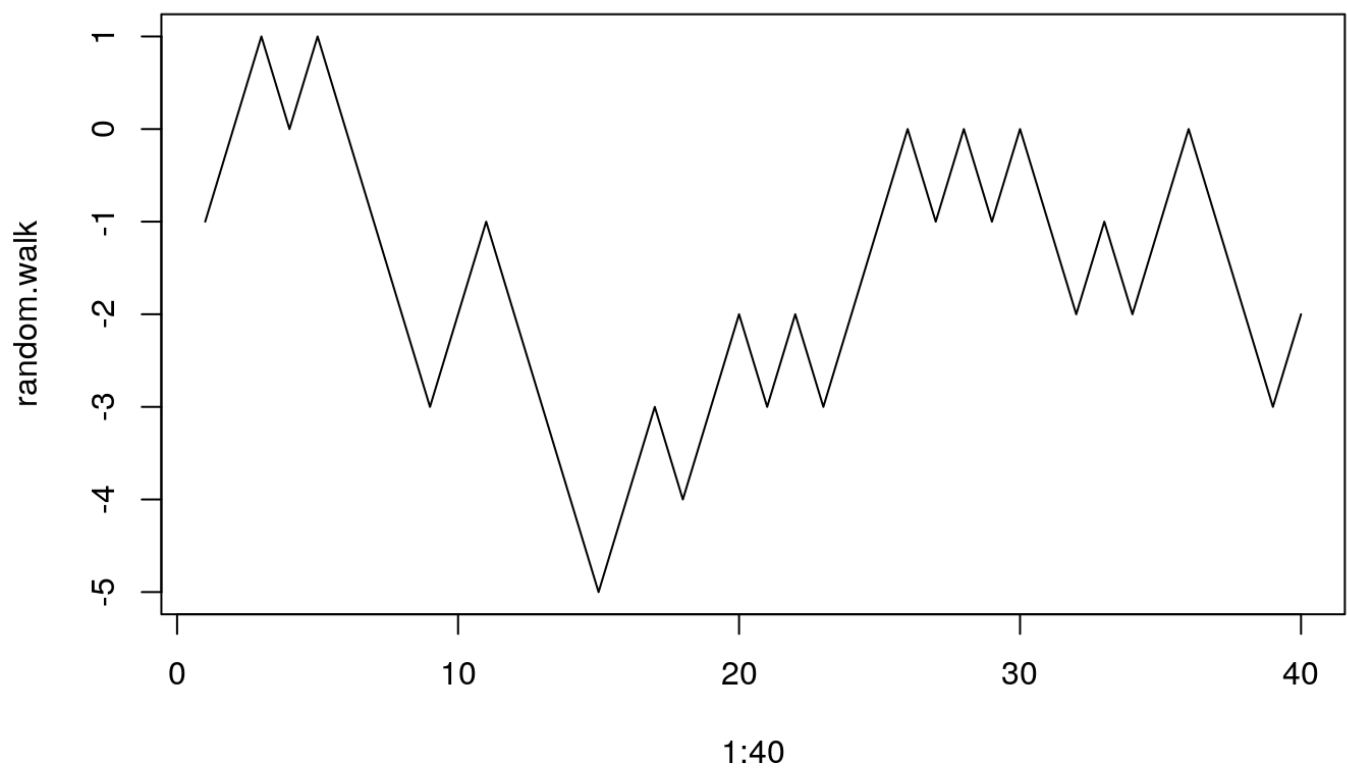
```
## here's the function
plot.random.walk <- function(seed)
{
  set.seed(seed)                                # this initializes the random
                                                # number generator used in
                                                # sample(). If we use the same
                                                # seed, we will get the same
                                                # set of "random" numbers.

  random.steps <- sample(x = c(-1,1), size = 40, replace = TRUE)
  ## this selects the number -1 or 1 randomly 40 times
  random.walk <- cumsum(random.steps)
  ## this turns the random steps into a walk by summing them up.
  ## cumsum(c(1,2,3)) returns a vector with values 1, 3, and 6.
  plot(1:40, random.walk, type = "l")
}
```

Here's an example

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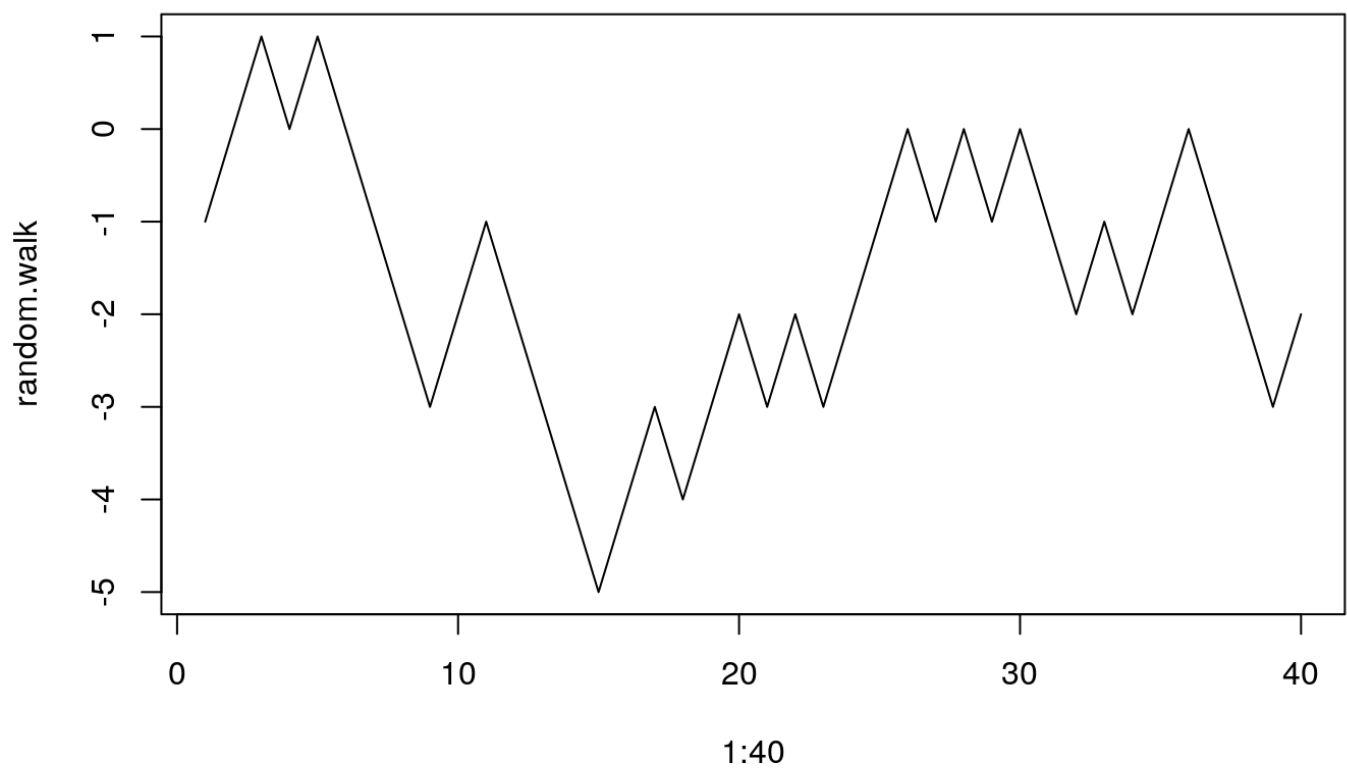
```
plot.random.walk(seed = 23)                    # Here we set seed to 23
```



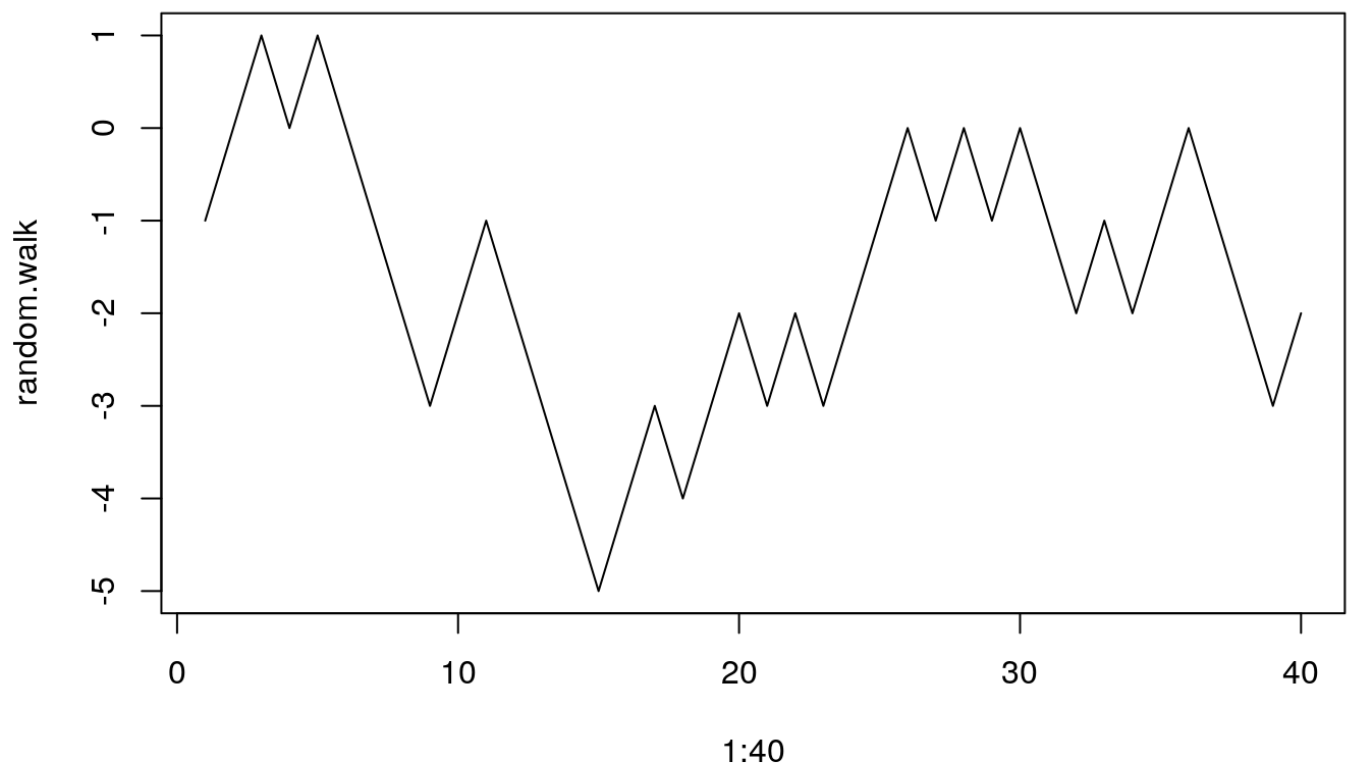
Modify the code below to try 1,2, 3, and 4 as “seeds”

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```
plot.random.walk(seed = 23)
```

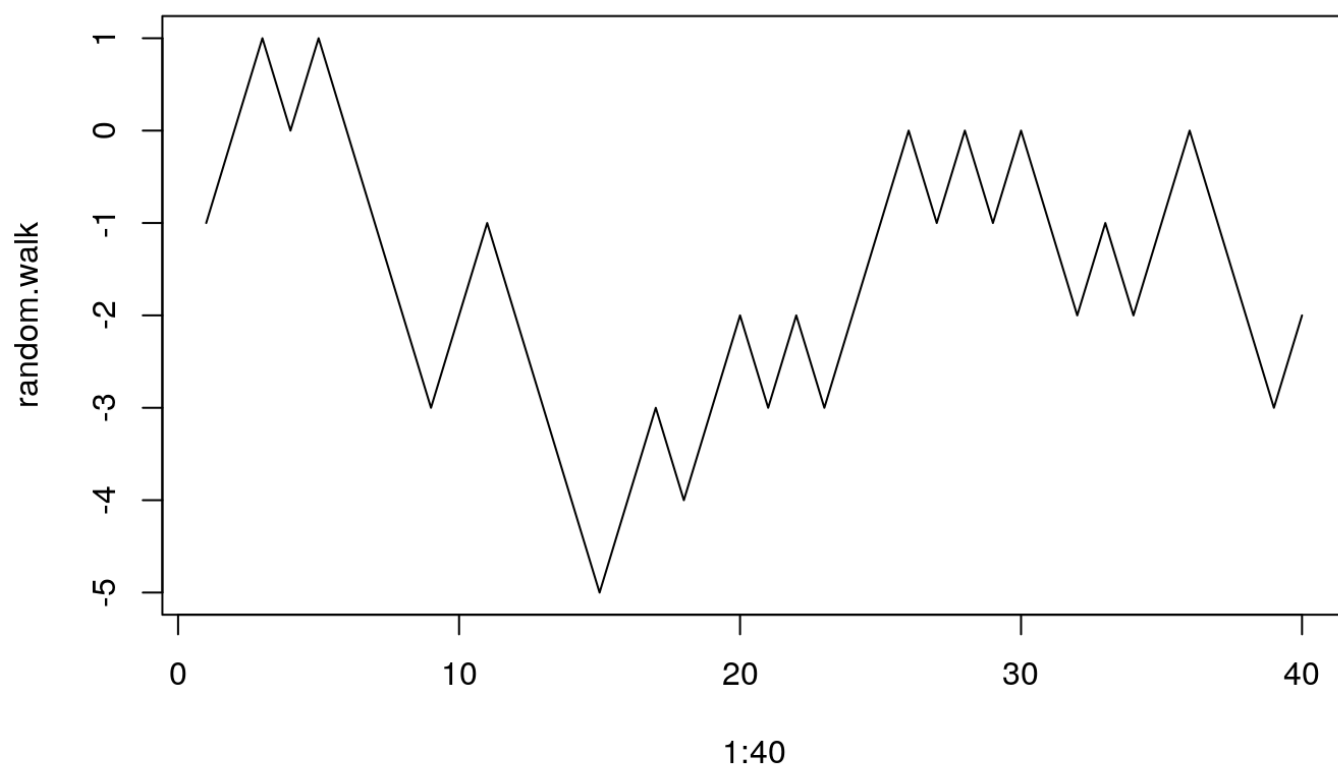
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```
plot.random.walk(seed = 23)
```



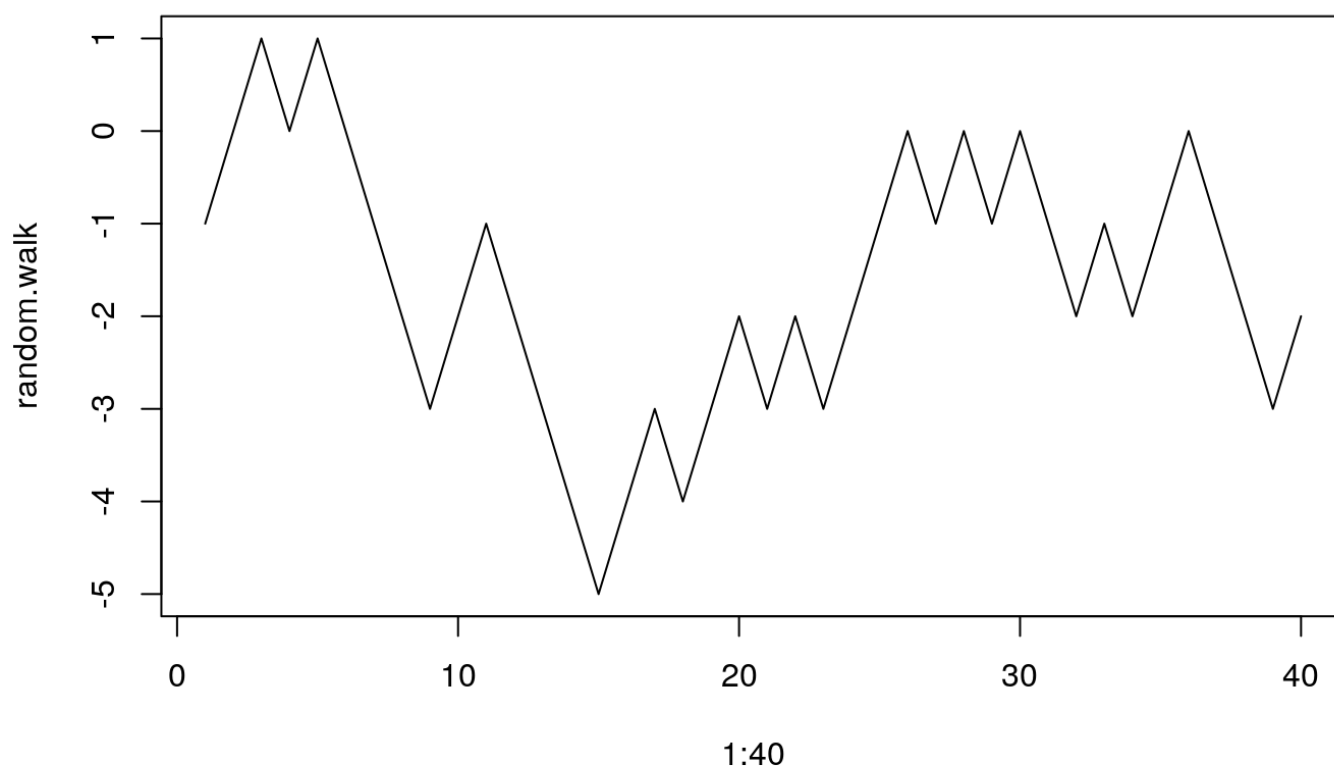
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```
plot.random.walk(seed = 23)
```



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```
plot.random.walk(seed = 23)
```

Q.2.2 Which of the following seems to be true

- A. Even if price changes are random, averaging zero, the realization of a random process can look like it has a trend.
- B. A historical trend is the one reality we observe and we shouldn't let anyone confuse us with alternative random realizations.
- C. Thinking about randomness may help us understand and interpret the reality we see and thus is a subject worth of study.

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```
## "Replace the NA with your answer (e.g., 'A' in quotes)"
answer2.2 = "C"
quiz.check(answer2.2)
```

Your answer2.2 : C

Warning in quiz.check(answer2.2) : Sorry, incorrect. Try again.

```
Warning in quiz.check(answer2.2) :
  Hint: Ignore automatic message about your answer being 'incorrect.'
  .There is no 'correct' answer. But of course a professor would lean
  .toward 'C'
```

Repeating for Tin

Plot the price of tin. You can use the same code we used for Chromium, modifying just slightly.

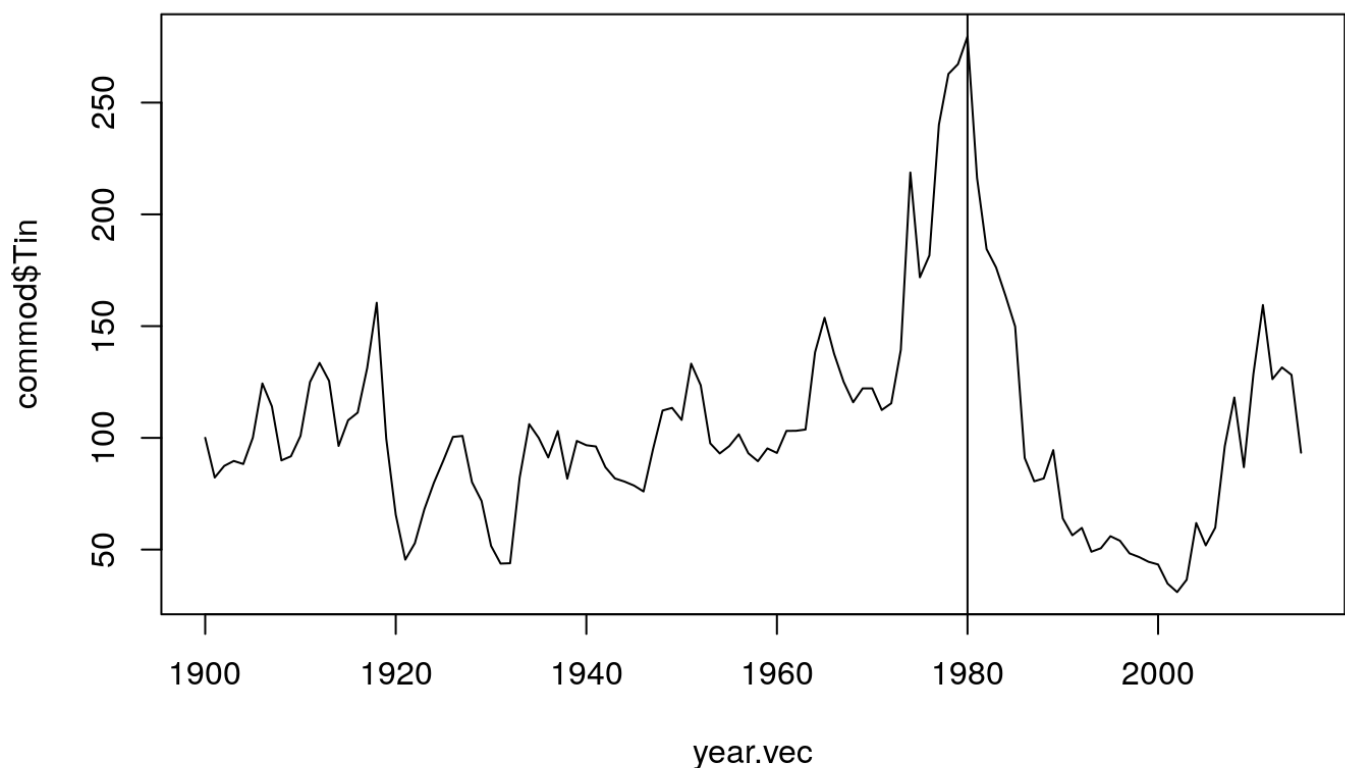
[Hide](#)

```
## put your code for Tin here  
## (Hint: you can copy and modify the commands we used for Chromium)  
commod["1963", "Tin"]
```

```
[1] 103.72
```

[Hide](#)

```
year.vec <- rownames(commod)  
plot(year.vec, commod$Tin, type = "l")  
abline(v = 1980)
```



Q2.3 Why do you think Ehrlich chose Tin as one of his commodities to bet on?

- A. Tin is difficult to substitute for
- B. Tin had shown a sharp increase in price for the decade or so before 1980 and so it looked like it would continue.
- C. The price had been going down before 1980 and was ready for a recovery
- D. There was evidence that the tin mines were running out of ore

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```
## "Replace the NA with your answer (e.g., 'A' in quotes)"
answer2.3 = "B"
quiz.check(answer2.3)
```

Your answer2.3 : B

Correct.

Explanation: Apparently, tin prices were controlled in part by a cartel of producers. See https://www.itri.co.uk/index.php?option=com_mtree&task=att_download&link_id=49605&cf_id=24 for a discussion from an industry research source.

Part 3. Replicating actual bet

In the original bet, Ehrlich was allowed to choose any commodities he wished. He chose Chromium, Copper, Tin, Tungsten, and Nickel. In our data, we don't have Tungsten. We can check to see if Simon still wins with the four commodities we have.

See the Lam reading, p. 1242 for an account of the bet.

We will use a function to compare commodities over time. This uses the indexing by label we did above. It returns the change in value of a \$1000 purchase of the commodity basket in the start year. So if the function returns +100, this means that the value of the commodity basket increased by \$100.

Q3.1 If the value of the basket increased by \$100, who would win?

A. Simon

B. Ehrlich

C. Neither, because the value of the basket doesn't tell us if *all* of the commodities increased in price at the same time.

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```
## "Replace the NA with your answer (e.g., 'A' in quotes)"
answer3.1 = "B"
quiz.check(answer3.1)
```

Your answer3.1 : B

Correct.

Explanation: An increase in prices suggests an increase in `.'scarcity'`, what Ehrlich was betting on.

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```
bet.fun <- function(start, end, basket.vec, data)
{
  ## start = 1980
  ## end = 1990
  ## data = commod
  ## basket.vec <- c("Copper", "Chromium", "Nickel", "Tin")
  year <- rownames(data)
  start.price <- data[year == start, basket.vec]
  end.price <- data[year == end, basket.vec]
  ## we invest 1000, buyin an equal $ amount in each commodity.
  start.funds <- 1000
  dollars.per.commod <- start.funds/length(basket.vec) # e.g. $250 with 4
  ## amount purchased of each commodity
  holdings.vec <- dollars.per.commod/start.price
  ## value at the end
  value.at.end <- sum(holdings.vec * end.price)
  change.in.value <- value.at.end - start.funds
  return(change.in.value)
}

who.wins.fun <- function(change.in.value)
{
  ifelse(change.in.value > 0, "ehrlich", "simon")
}
```

Let's try this for 1980 to 1990 with Ehrlich's basket.

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```
ehrlich.basket.vec <- c("Copper", "Chromium", "Nickel", "Tin")
original.bet.result <- bet.fun("start" = 1980,
                              "end" = 1990,
                              "basket.vec" = ehrlich.basket.vec,
                              "data" = commod)
who.wins.fun(original.bet.result)
```

```
[1] "simon"
```

Q3.2 How much would Simon have won with our version of the “bet”?

- A. About 300 dollars
- B. He wouldn't have won. He would have lost about 300 dollars
- C. About 500 dollars
- D. About 100 dollars

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```
## "Replace the NA with your answer (e.g., 'A' in quotes)"
answer3.2 = "A"
quiz.check(answer3.2)
```

Your answer3.2 : A

Correct.

Explanation: A negative number means prices went down, and so Simon would have won.

Part 4. Does Simon win in other periods?

Commodities are highly volatile. Even if commodities were generally rising, Ehrlich could have been unlucky about the timing of his bet.

Let's check over all years since 1900. We can construct a "loop" using R that will repeat the bet according to our instructions. We can then save the output in a vector and make a judgement about whether Ehrlich was simply unlucky.

Here's a simple example of a loop, in which we repeat an action 7 times, incrementing the value of "i", and printing this value along with a label "iteration".

[Hide](#)

```
for (i in 1:7)  ## the variable "i" takes the value 1, 2, 3, ...
{
  ## anything in between {} is done once
  ## for each value that "i" takes
  print(c("iteration", i))
}
```

```
[1] "iteration" "1"
[1] "iteration" "2"
[1] "iteration" "3"
[1] "iteration" "4"
[1] "iteration" "5"
[1] "iteration" "6"
[1] "iteration" "7"
```

Now let's do a loop to see who won the bet in every year of the last century.

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```
start.vec <- 1900:2005      ## a vector of years to start the bet
bet.result.vec <- rep(NA, length(start.vec)) ## an empty vector to be
                                ## used to store the
                                ## results
names(bet.result.vec) <- start.vec ## labeling the elements (you can ignore)
for (i in 1:length(start.vec))
{
  bet.result.vec[i] <- bet.fun(start = start.vec[i],
                              end = start.vec[i] + 10,
                              basket.vec = ehrlich.basket.vec,
                              data = commod)
}
print(bet.result.vec)
```

1900	1901	1902	1903	1904	1905	1906	
1907	1908	1909	1910				
-258.700000	-156.431856	49.123809	-2.021327	-127.212088	-95.459327	-52.223399	-2
1.845369	293.912953	-175.917402	-424.804605				
1911	1912	1913	1914	1915	1916	1917	
1918	1919	1920	1921				
-552.181803	-531.713958	-479.901991	-406.828864	-457.454997	-483.863734	-488.154673	-50
1.451139	-259.693188	-81.082163	216.717344				
1922	1923	1924	1925	1926	1927	1928	
1929	1930	1931	1932				
243.454453	252.717049	341.054540	365.566495	331.228393	386.168194	276.108124	30
6.675598	392.074108	414.174885	351.395641				
1933	1934	1935	1936	1937	1938	1939	
1940	1941	1942	1943				
38.681985	-29.656863	-72.914365	-150.671378	-113.895743	74.669806	-21.202224	-
3.043024	132.044717	202.843571	269.671205				
1944	1945	1946	1947	1948	1949	1950	
1951	1952	1953	1954				
213.919871	395.367748	542.721218	291.148075	176.863221	283.016148	92.501372	-3
4.354117	84.176895	16.043674	199.861978				
1955	1956	1957	1958	1959	1960	1961	
1962	1963	1964	1965				
170.674010	38.283426	107.300058	183.451841	109.764135	286.654596	400.477990	18
1.086693	314.033040	526.655925	591.149632				
1966	1967	1968	1969	1970	1971	1972	
1973	1974	1975	1976				
599.870085	591.992619	507.597248	647.651367	627.592390	308.600045	97.417259	-3
8.528232	-294.608549	-353.756361	-475.654892				
1977	1978	1979	1980	1981	1982	1983	
1984	1985	1986	1987				
-441.250784	71.947758	-57.956949	-306.220473	-275.830073	-193.578232	-353.610315	-23
7.373103	35.802145	16.675437	-96.471592				
1988	1989	1990	1991	1992	1993	1994	
1995	1996	1997	1998				
-524.143483	-588.938139	-354.297500	-393.554388	-372.843395	-10.989176	366.094672	6
8.122978	768.730531	1523.535732	2017.405153				
1999	2000	2001	2002	2003	2004	2005	
1150.647495	1747.100492	2599.771715	2193.400721	1525.633810	609.806673	216.228316	

To summarize our results, let's look at them in several ways.

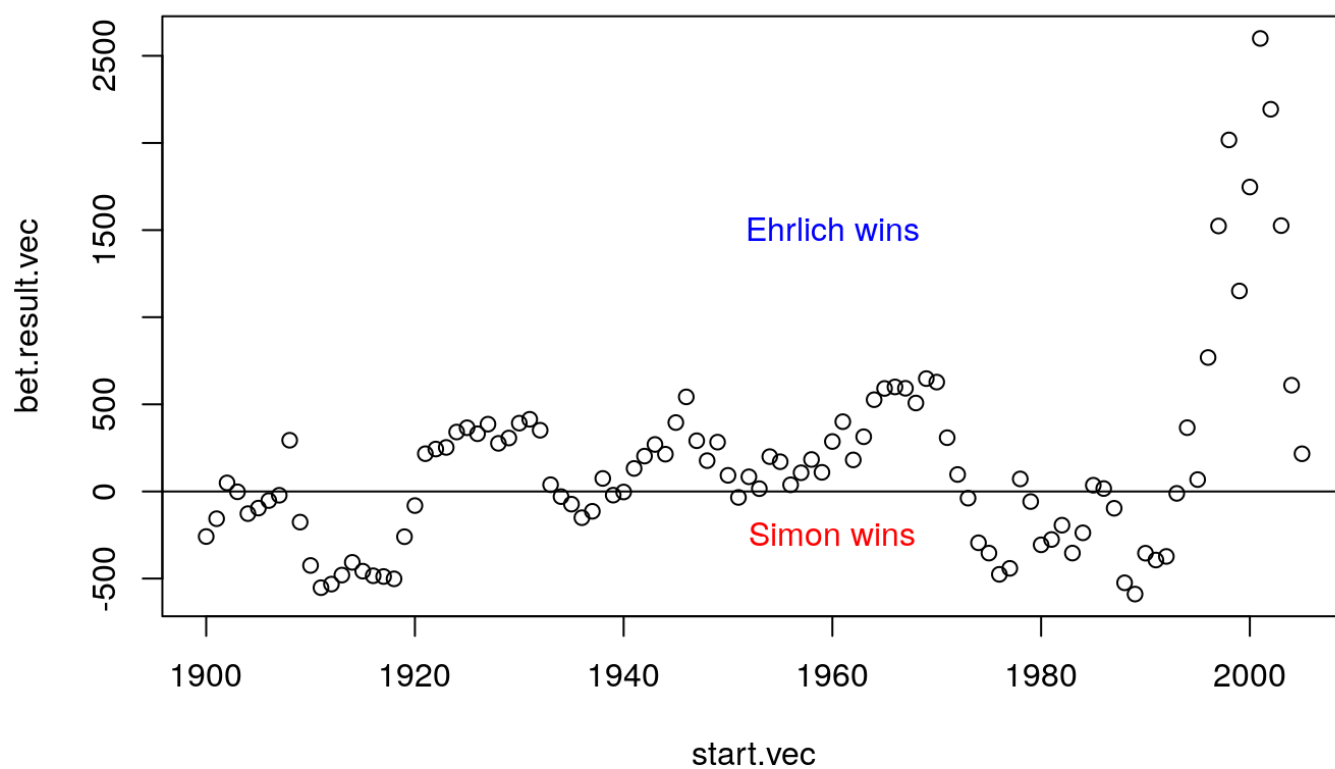
Plot the numbers

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```
plot(start.vec, bet.result.vec)
abline(h = 0)
```

Hide

```
text(1960, 1500, "Ehrlich wins", col = "blue")
text(1960, -250, "Simon wins", col = "red")
```



In the graph, it looks like Ehrlich would have won more of the time. But not a lot more.

Let's tabulate how many times Ehrlich and Simon would have won:

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```
## Convert to a vector of "simon" and "ehrlich", depending on who won
winners.vec <- who.wins.fun(bet.result.vec)
print(winners.vec)
```

	1900	1901	1902	1903	1904	1905	1906	1907	1908
8	1909	1910	1911	1912					
	"simon"	"simon"	"ehrlich"	"simon"	"simon"	"simon"	"simon"	"simon"	"ehrlich"
	"simon"	"simon"	"simon"	"simon"					
	1913	1914	1915	1916	1917	1918	1919	1920	1921
1	1922	1923	1924	1925					
	"simon"	"simon"	"simon"	"simon"	"simon"	"simon"	"simon"	"simon"	"ehrlich"
	"ehrlich"	"ehrlich"	"ehrlich"	"ehrlich"					
	1926	1927	1928	1929	1930	1931	1932	1933	1934
4	1935	1936	1937	1938					
	"ehrlich"	"ehrlich"	"ehrlich"	"ehrlich"	"ehrlich"	"ehrlich"	"ehrlich"	"ehrlich"	"simon"
	"simon"	"simon"	"simon"	"ehrlich"					
	1939	1940	1941	1942	1943	1944	1945	1946	1947
7	1948	1949	1950	1951					
	"simon"	"simon"	"ehrlich"	"ehrlich"	"ehrlich"	"ehrlich"	"ehrlich"	"ehrlich"	"ehrlich"
	"ehrlich"	"ehrlich"	"ehrlich"	"simon"					
	1952	1953	1954	1955	1956	1957	1958	1959	1960
0	1961	1962	1963	1964					
	"ehrlich"	"ehrlich"	"ehrlich"	"ehrlich"	"ehrlich"	"ehrlich"	"ehrlich"	"ehrlich"	"ehrlich"
	"ehrlich"	"ehrlich"	"ehrlich"	"ehrlich"					
	1965	1966	1967	1968	1969	1970	1971	1972	1973
3	1974	1975	1976	1977					
	"ehrlich"	"ehrlich"	"ehrlich"	"ehrlich"	"ehrlich"	"ehrlich"	"ehrlich"	"ehrlich"	"simon"
	"simon"	"simon"	"simon"	"simon"					
	1978	1979	1980	1981	1982	1983	1984	1985	1986
6	1987	1988	1989	1990					
	"ehrlich"	"simon"	"simon"	"simon"	"simon"	"simon"	"simon"	"ehrlich"	"ehrlich"
	"simon"	"simon"	"simon"	"simon"					
	1991	1992	1993	1994	1995	1996	1997	1998	1999
9	2000	2001	2002	2003					
	"simon"	"simon"	"simon"	"ehrlich"	"ehrlich"	"ehrlich"	"ehrlich"	"ehrlich"	"ehrlich"
	"ehrlich"	"ehrlich"	"ehrlich"	"ehrlich"					
	2004	2005							
	"ehrlich"	"ehrlich"							

Hide

```
## count number of times each won using the "table()" function
table(winners.vec)
```

```
winners.vec
ehrlich  simon
    62     44
```

Q4.1 What percent of the time would Ehrlich have won of the 106 simulated bets from 1900 to 2005?

- A. About 62 percent
- B. About 58 percent

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```
## "Replace the NA with your answer (e.g., 'A' in quotes)"
answer4.1 = "B"
quiz.check(answer4.1)
```

Your answer4.1 : B

Correct.

Explanation: 62/106 is about 0.58.

Make a fancier plot. (You don't need to understand this code.)

Hide

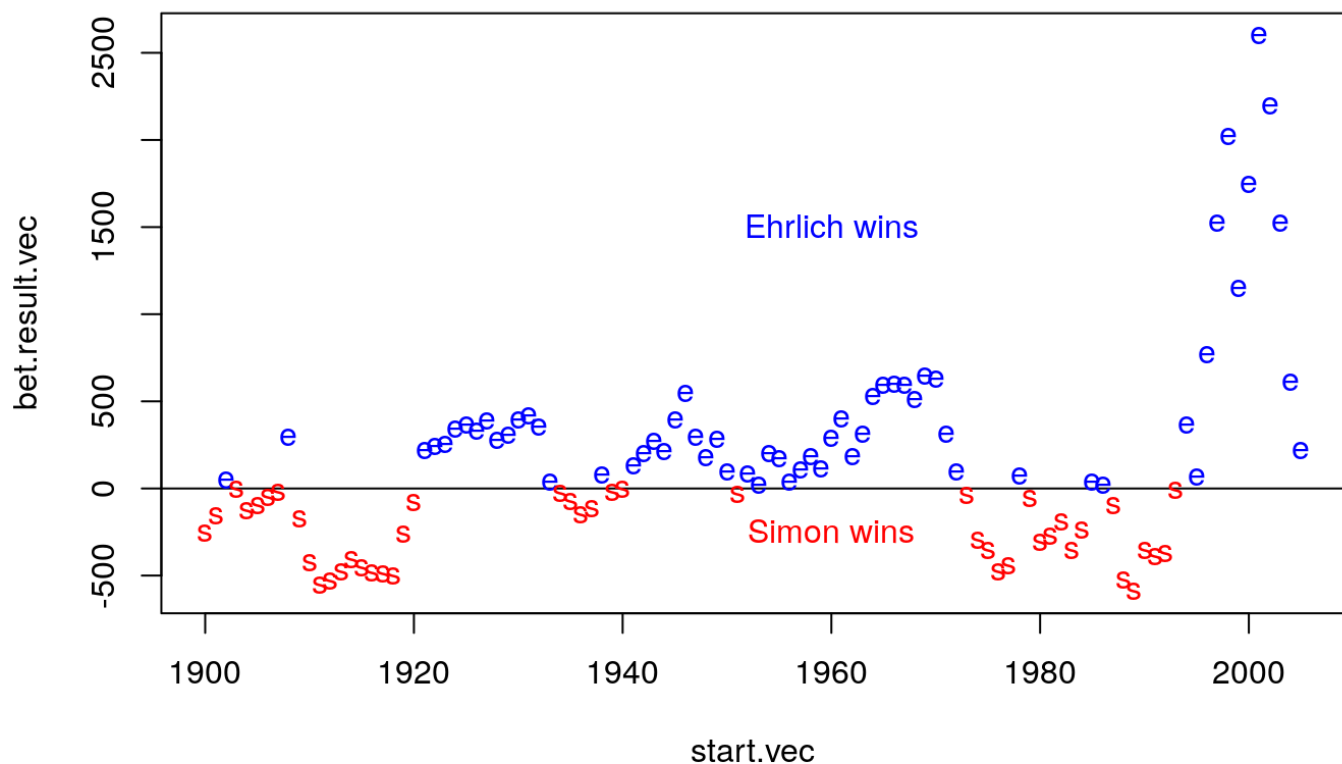
```
plot(start.vec, bet.result.vec, type = "n") #an empty plot
abline(h = 0)
```

Hide

```
e <- which(winners.vec == "ehrlich")    # T or F, to subset text()
s <- which(winners.vec == "simon")
text(x = start.vec[e], y = bet.result.vec[e], labels = "e", col = "blue")
text(x = start.vec[s], y = bet.result.vec[s], labels = "s", col = "red")
```

Hide

```
text(1960, 1500, "Ehrlich wins", col = "blue")
text(1960, -250, "Simon wins", col = "red")
```



5. Generalizing the bet to other commodities

A change in the supply of a particular commodity may or may not cause a large price change – depending on the availability of substitutes. (Note: the results from this section will be useful for your write-up of the graded questions at the end of the lab.)

The following code provides a template for a single commodity, in this case “Chromium”.

[Hide](#)

```
start.vec <- 1900:2005
result.vec <- rep(NA, length(start.vec))
for (i in 1:length(start.vec))
{
  result.vec[i] <- bet.fun(start = start.vec[i],
                        end = start.vec[i] + 10,
                        basket.vec = c("Chromium"),
                        data = commod)
}
winner.vec <- who.wins.fun(result.vec)
cat("counts:\n")
```

```
counts:
```

[Hide](#)

```
table(winner.vec)
```

```
winner.vec
ehrllich  simon
      61      45
```

[Hide](#)

```
cat("proportions:\n")
```

```
proportions:
```

[Hide](#)

```
prop.table(table(winner.vec))
```

```
winner.vec
ehrllich  simon
0.5754717 0.4245283
```

Well done. You are finished with the computing portion of Lab 4.

Part 5. Graded Questions

1. What relationship did Ehrlich expect to hold between commodity prices and population growth?

Ehrlich was worried that the rapid population growth could lead to resource starvation. Due to that, he was worried that inflation would be a problem.

2. Who won the bet in 1990? Given our analysis of other time periods, would you say that the winner was “lucky” or “right”, or both? [Explain your answer in 50 to 100 words.]

As noted in the lab, this section is for you to explore and describe what you found. Whether what you find is conclusive or inconclusive, you can still get full credit by reporting whatever you found.

Simon won the bet in 1990. Simon was lucky, because according to the statistics Ehrlich had a 57.55% chance of winning from 1900 to 2000s.

3. Choose another easily substitutable commodity from the database. Conduct the analysis of its price as the previous part of the lab and answer the following questions.

- 3.1) Why do you think this commodity would be easily substitutable, explaining what “service” it provides with 1 or 2 sentences.

Beef, as it provides a source of protein. If beef is scarce, then it can be easily replaced with other proteins such as chicken or pork. Newer advancements in food technology has allowed for plant based substitution.

- 3.2) Attach a plot of the change of prices (“bet.result”) of the chosen commodity.

Hint 1: there would only be your chosen commodity in the “basket.vec”. You can either modify the original “ehrllich.basket.vec” or create your own “basket.vec”. Remember to modify for-loop consistently.

Hint 2: you don't need to show whether Ehrlich or Simon win about this commodity. But labelling the plot with different colors can help you understand the general trend.

Hint 3: NOT REQUIRED. To make it a better graph, you can add a title to the graph indicating the commodity you choose adding 'main = "commodity name"' to the syntax.

Hide

```
brian.basket.vec <- c("Beef")
start.vec <- 1900:2005      ## a vector of years to start the bet
bet.result.vec <- rep(NA, length(start.vec)) ## an empty vector to be
                                     ## used to store the
                                     ## results
names(bet.result.vec) <- start.vec ## labeling the elements (you can ignore)
for (i in 1:length(start.vec))
{
  bet.result.vec[i] <- bet.fun(start = start.vec[i],
                              end = start.vec[i] + 10,
                              basket.vec = brian.basket.vec,
                              data = commod)
}

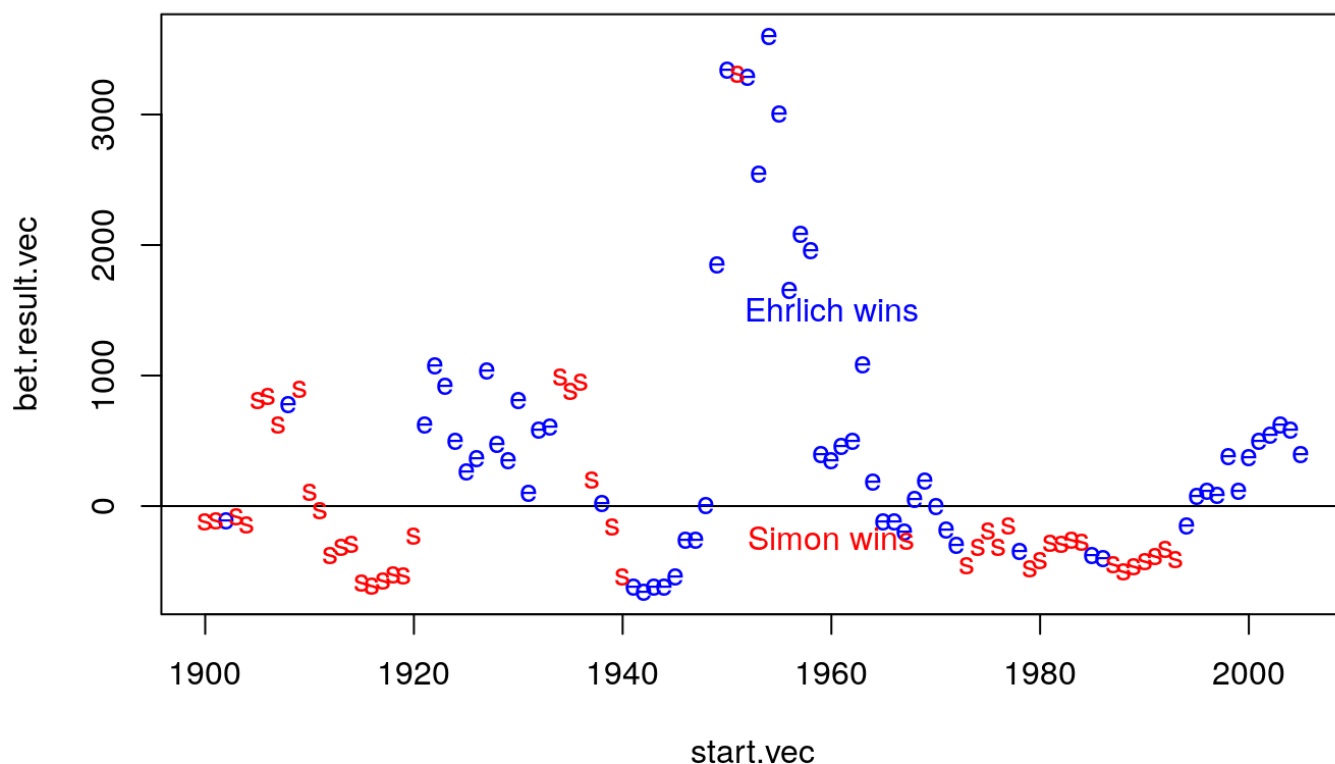
plot(start.vec, bet.result.vec, type = "n") #an empty plot
abline(h = 0)
```

Hide

```
e <- which(winners.vec == "ehrlich") # T or F, to subset text()
s <- which(winners.vec == "simon")
text(x = start.vec[e], y = bet.result.vec[e], labels = "e", col = "blue")
text(x = start.vec[s], y = bet.result.vec[s], labels = "s", col = "red")
```

Hide

```
text(1960, 1500, "Ehrlich wins", col = "blue")
text(1960, -250, "Simon wins", col = "red")
```



4. Choose a commodity from the database which you think would be hard to substitute for. And explain why you think this commodity would NOT be easily substituted, explaining what “service” it provides. (In 1 or 2 sentences) [Note: you don’t need to analyze the data for this problem.]

Iron is a commodity that is probably hard to substitute for. It’s one of the strongest metals used for construction. Without iron, we could probably not make the most sustainable and sturdiest buildings, such as workplaces or housing, which is the service it provides.

5. What are two reasons prices might not reflect the long-term availability of a commodity? [A sentence on each.]

One might be due to a trend in consumer products, maybe a product or commodity has gone viral and has increased suddenly. Another might be due to epidemic or natural disasters, where examples include panic buying from COVID-19 or grocery stock shortages during natural disasters.

6. For the following questions, use the “app” available <http://shiny.demog.berkeley.edu/josh/harvest/> (http://shiny.demog.berkeley.edu/josh/harvest/) to gather data on the relationship between harvest rate “h” and the sustainable (steady-state) yield. You can do this by moving the slider from 0 upwards by increments of 0.02.

6.1 There is a slight bug in the app, so that it reports sustainable yields even when harvest rate is unsustainable. How can you inspect the graphs to check on sustainability?

You can manually inspect the graph by observing that the point where the green curve intersects the red curves. When the red curve is above the green curve, this shouldn't be sustainable.

6.2 When $h = 0.01$, the sustainable yield is 0.0678, what is the sustainable yield when $h = 0.02$ and $h = 0.04$?

When $h = 0.02$, the sustainable yield is 0.108; $h = 0.04$, the sustainability yield is 0 (this is unsustainable due to the bug).

6.3 What harvest rate maximizes the sustainable yield? [Hint: try all of the slider values]

At $h=0.022$, the sustainable yield is maximized.

6.4 What harvest rate maximizes the yield during the first year of harvesting (this is about year 10 on the slider, or year 0 on the time axes of the graph)? (You can either report the value you find with the “app” or your theoretical answer)

With the app, the answer is $h=0.15$. But assuming it can go higher, 1 would be the theoretical peak.

6.5 Would harvest rates be higher in year “1” if we harvested at the sustainable rate found in part 6.3, or if they were .02 higher than the sustainable rate?

The sustainable yield would be higher at $h = 0.024$. But we would get diminishing returns in the long run.

6.6 When Becker advocates for attention to sustainability, how did Lam reply to his warning? (In 1 or 2 sentences).

Lam replies that instead of attention to sustainability, we need better institutions to manage the sustainability, such as functioning markets.

Congratulations! You are finished with Lab 4.