

PHP 2500 Introduction to Biostatistics

Problem Set One Solutions (Updated)

1. (Pagano #7, p30) (a) discrete, (b) continuous, (c) continuous, (d) discrete

(Pagano #9, 15, p30) self explanatory

2. (Pagano #8, p30)

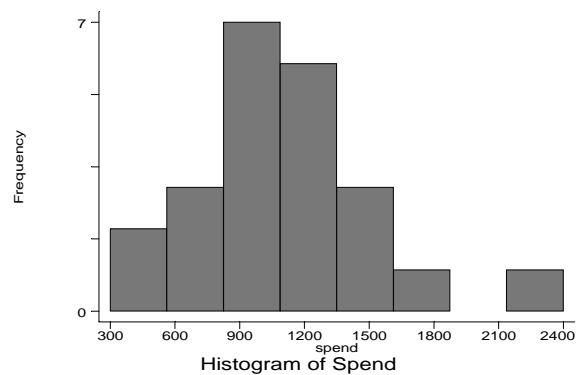
(a) Sort by Per Capita Expenditure:

```
. sort spend
. list
    country    spend
1.    Greece    371
2.    Portugal  464
3.    Spain     644
4.    Ireland   658
5.    New Zealand 820
6.    Britain   836
7.    Denmark   912
8.    Belgium   980
9.    Australia 1032
10.   Japan     1035
11.   Italy     1050
12.   Finland   1067
13.   Austria   1093
14.   Netherlands 1135
15.   Luxembourg 1193
16.   Germany   1232
17.   Norway    1234
18.   France    1274
19.   Iceland   1353
20.   Sweden    1361
21.   Switzerland 1376
22.   Canada    1683
23.   United states 2354

. gsort - spend
. list
    country    spend
1. United states 2354
2.    Canada    1683
3.    Switzerland 1376
4.    Sweden     1361
5.    Iceland    1353
6.    France     1274
7.    Norway     1234
8.    Germany    1232
9.    Luxembourg 1193
10.   Netherlands 1135
11.    Austria    1093
12.   Finland    1067
13.    Italy      1050
14.    Japan      1035
15.   Australia   1032
16.    Belgium    980
17.    Denmark    912
18.    Britain    836
19.   New Zealand 820
20.    Ireland    658
21.    Spain      644
22.    Portugal   464
23.    Greece     371
```

(b) Histogram

```
. hist spend, frequency bin(8) xlabel(300(300)2400) title ("Histogram of Spend")
```



2. (c) The shape of the histogram is fairly 'bell-shaped', with maybe an extended tail towards higher per Capita Expenditure. Note that the shape of your histogram will depend on the number of bins.

(d) Mean = 1093.783, Variance (S.D.) = 170,223.41 (412.58)
Range = 1983, Median 1067.

3. (Pagano #6, p59)

(a) Mean = 25.9, Median = 24, Modes(s) = 12 & 24, Range = 95.9,
IQR(stata) = 39 - 2.25 = 36.75, IQR(us) = 36 - 4 = 32,
Standard deviation = 27.4

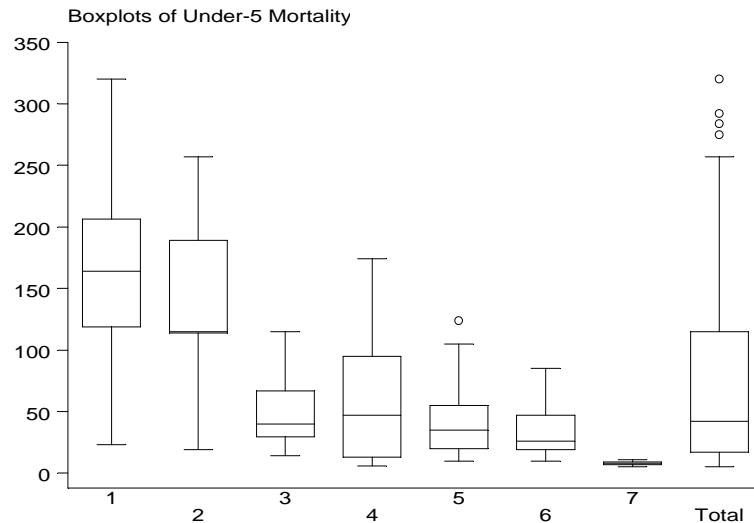
- (b) You can verify that if you subtract the mean, $\bar{x} = 25.9115$, from each of these thirteen numbers, the results will sum to zero. Or you can use algebra:

$$\begin{aligned}\sum_{i=1}^{13} (x_i - \bar{x}) &= \sum_{i=1}^{13} x_i - \sum_{i=1}^{13} \bar{x} \\ &= \sum_{i=1}^{13} x_i - 13\bar{x} \\ &= \sum_{i=1}^{13} x_i \\ &= 13 \frac{i=1}{13} - 13\bar{x} = 13\bar{x} - 13\bar{x} = 0\end{aligned}$$

4. (a) The Mean daily rainfall in 1999 was probably greater in Providence than in Tucson.
- (b) The mode daily rainfall in 1999 for both cities is zero. The most frequent amount of rain is none even though the exact number of days without rain could be different between the two cities.
- (c) The median daily rainfall in 1999 for both cities is zero. This is because it rains less than half of the days during the year. Thus, the 50% percentile has to be zero.
5. (a) Smallest mean -- Europe (averaging smaller numbers)
Largest median -- Africa (histogram shows 50% is greatest)
Smallest Standard deviation -- Europe (smallest spread in data)
- (b) The mean and median are nearly equal when the data are distributed nearly symmetrically, as they are in Africa. They differ when the numbers are skewed, as they are in Asia, because the mean depends heavily on outliers. For Asia, the median will be less than the mean because of the skewness.

6. (a)

```
. graph box mortality, over(region, total label(alternate)) cap(10)
title(Boxplots of Under-5 Mortality) ylab(0(50)350)
```



Key:

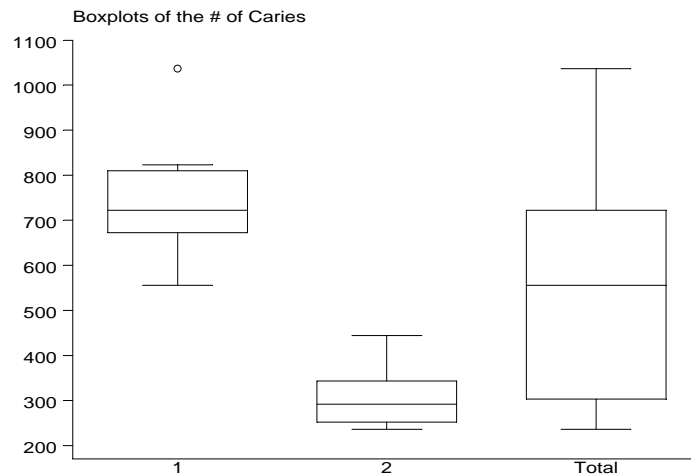
- 1) Sub-Saharan Africa
- 2) North Africa & Middle East
- 3) South Asia
- 4) East Asia and Pacific
- 5) Latin America and Caribbean
- 6) Europe, Commonwealth, and Baltic
- 7) Industrial

(a) Notice that groups 1 and 2 have similarly high Under5-mortality with a wide spread, while groups 3,4,5, and 6 all tend to be more compact and less variable. Finally, the industrial nations tend to have little variability and a low Under-5 mortality.

- (b) (i) Sub-Saharan
(LH =128, M=164, UH =203, IQR=75, LF =15.5, UF =315.5)
Niger is an in the Sub-Saharan.
- (ii) Middle East
(LH =34, M=40, UH =63, IQR=29, LF =0, UF =106.5)
Sudan and Yemen are outliers in the Middle east

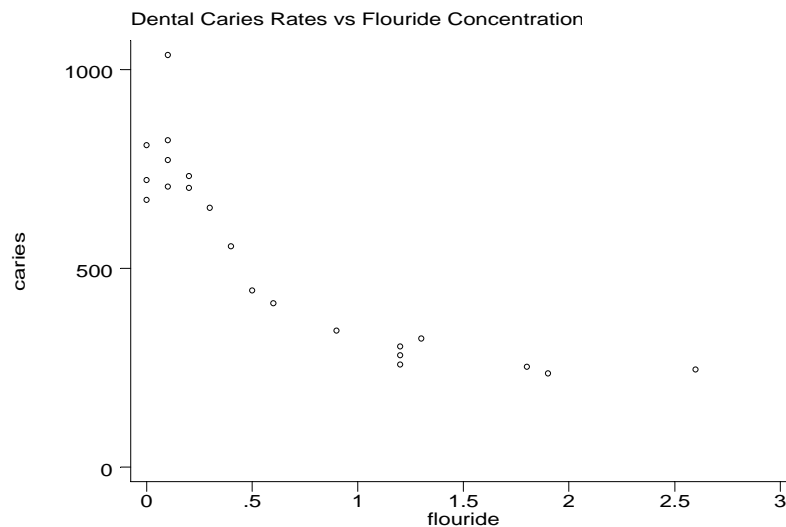
7. (a)

```
. graph box caries, over(group, total) title(Boxplots of the # of
Caries) ylab(200(100)1100)
```



(b)

```
. graph twoway scatter caries fluoride, title(Dental Caries Rates vs
Flouride Concentration) ylab(0(100)1000) xlab(0(0.5)3) msymbol(oh)
```



- (c) About 250 carries per 100 children, or 2.5 per child.
 - (d) It looks like an increase in fluoride concentration from 0.25 to 1.25 will cut the caries rate nearly in half.
 - (e) An increase in fluoride concentration from 2.0 to 3.0 would appear to have little effect.
8. (a) Crude Birth rate: $87,202/6,060,943 = 0.0144$ 14.39 per 1000
 (b) Crude death rate: $53,804/6,060,943 = 0.0089$ 8.88 per 1000
 (c) Infant Mortality rate: $596/87,202 = 0.0068$ 6.83 per 1000
9. The death rate from all causes must have been much greater in Guatemala City than in Lima. Specifically, from the facts that were given, we can calculate that the death rate from all causes in Guatemala City was 12.9 per 1000, while in Lima it was only 5.4 per 1,000, so that the Guatemala City rate was more than double the Lima rate ($12.9/5.4 = 2.39$).

The calculation goes like this: We're given that for Guatemala City

$$\% \text{ of all deaths due to cancer} = \frac{\text{CancerDeaths}(GC)}{\text{TotalDeaths}(GC)} = 0.135 \quad (13.5\%)$$

$$\text{death rate for cancer} = \frac{\text{CancerDeaths}(GC)}{\text{Popultaion}(GC)} = 0.001736 \quad (173.6/100,000)$$

Dividing the second by the first, we find that:

$$\text{All-cause DR} = \frac{\text{TotalDeaths}(GC)}{\text{Popultaion}(GC)} = 0.001736 / 0.135 = 1285.9 / 100,000$$

or 12.9 per 1000.

A similar calculation shows that the rate in Lima was 5.4 per 1000.

- 10.(a) The first explanation that comes to mind is that the quality of care has improved. Later you might wonder if the population of babies has changed, which might lead you to do the analysis in (b).
- (b) The weight-specific mortality rates (per 100 babies, in order of ascending Birth Weight categories) in 1953-57 were

92.1, 42.9, 12.9, and 6.0, and the corresponding rates in 1968-72 were 91.8, 55.7, 13.7, and 4.0.

These changes cannot explain the 40% drop in the crude mortality rate. If we adjust to the 1953-57 weight distribution, we find that the adjusted rate for 1968-72 is $[(0.918)(126) + (0.557)(112) + (0.435)(241) + (0.040)(601)]/1080 = 0.284$, or **28.4** deaths per 100 babies, slightly greater than the 1953-57 rate (which was 21.4 per 100).

- (c) The decline in the crude mortality rate can be explained entirely by the changes in the weight distribution. (There were relatively small changes in the weight-specific weights, and those changes were not all in the same direction.) The mortality rate in the smallest weight group was just as high in 1970 (92 per 1000) as it was in 1955. But such babies made up a much smaller proportion of the population in 1980, and that is what brought the crude rate down.
- (d) $SMR = (\text{observed deaths})/(\text{Expected deaths}) = \text{crude}/\text{IA rates}$.
 $SMR(1953-57) = 0.2139/0.2839 = 0.75$, $SMR(1968-72) = 1$

```
. dstsize death number weight_ca, by(year) base(1953)
```

```
-> year= 1953
```

Stratum	Pop.	Cases	Unadjusted Pop. Dist.	Unadjusted Rate[s]	Std. Pop. Dst[P]	s*P
1	126	116	0.117	0.9206	0.117	0.1074
2	112	48	0.104	0.4286	0.104	0.0444
3	241	31	0.223	0.1286	0.223	0.0287
4	601	36	0.556	0.0599	0.556	0.0333
Totals:	1080	231	Adjusted Cases:		231.0	
			Crude Rate:		0.2139	
			Adjusted Rate:		0.2139	
			95% Conf. Interval: [0.1960, 0.2318]			

```
-> year= 1968
```

Stratum	Pop.	Cases	Unadjusted Pop. Dist.	Unadjusted Rate[s]	Std. Pop. Dst[P]	s*P
1	49	45	0.040	0.9184	0.117	0.1071
2	106	59	0.086	0.5566	0.104	0.0577
3	92	40	0.075	0.4348	0.223	0.0970
4	985	39	0.800	0.0396	0.556	0.0220
Totals:	1232	183	Adjusted Cases:		349.8	

Crude Rate: 0.1485
Adjusted Rate: 0.2839
95% Conf. Interval: [0.2568, 0.3110]

Summary of Study Populations:

year	N	Crude	Adj_Rate	Confidence Interval	
1953	1080	0.213889	0.213889	[0.195970,	0.231808]
1968	1232	0.148539	0.283919	[0.256844,	0.310994]

. di 0.2139/0.2839
.75343431

- (e) If we adjust to the 1968-72 rates, we find that the adjusted rate for 1953-57 is
 $[(0.918)(126) + (0.557)(112) + (0.435)(241) + (0.040)(601)]/1080$
= **0.284**, or **28.4** deaths per 100 babies,
slightly greater than the 1953-57 rate (which was 21.4 per 100).
This is the same as (b).

11. (a) 1940: $15820/131670 = 1.201$ per 1000
1986: $469330/241097 = 1.947$ per 1000
 $(1.947/1.201 = 1.62$; 62% greater in 1986)

- (b) 1986 population concentrated more in older age groups.

- (c) There is a relationship – the death rate rises sharply in the older age groups.

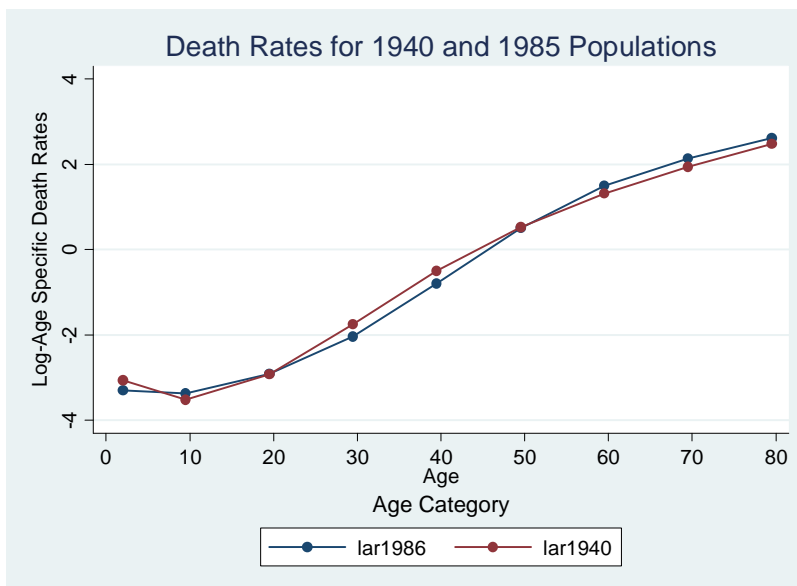
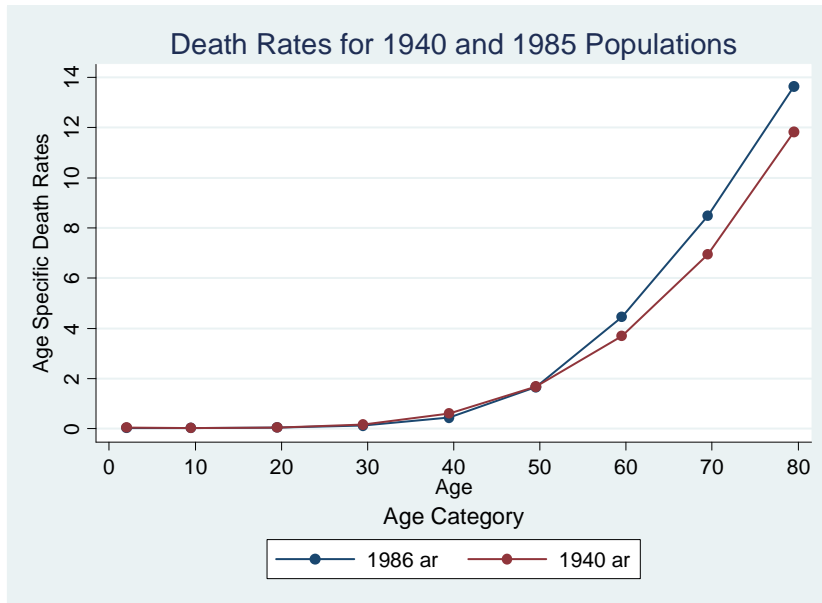
11. (d) Yes, because (i) rate changes with age, and (ii) the two populations have somewhat different age distributions.

- (e) Age-adjusted rates: 1940 1.201 per 1000 (= crude rate)
1986 1.339 per 1000
(now only 11% greater than 1940)

- (f) 1940 Age-adjusted rate = Crude
(because 1940 population was used as the standard)
1986 Age adjusted rate < Crude
(because rate increase with age, and 1940 population used as standard has “younger” age distribution.)

- (g) The attached graphs show that the age-specific rate in the two populations follow the same overall pattern, increasing rapidly with age. For this reason it is appropriate to adjust. This is a situation where all “reasonable” choices of a standard population will give a ratio of adjusted rates (1986/1940) that is greater than 1, but much smaller than the ratio of crude rates. For example, we saw above that with the 1940 population as the standard, the

ratio of adjusted rates is 1.11. If instead we use 1986 as the standard, the ratio of adjusted rates is 1.14. Both show 1986 is 10-15% higher, and both are much less than the ratio of crude rates (part (a)), where 1986 was 62% higher than 1940.



Graph one Stata commands:


```
. graph twoway connected ar1986 Age || connected ar1940 Age,
    ylab(0(2)14) xlab(0(10)80) title("Death Rates for 1940 and 1986
    Populations") l2("Age Specific Death Rates") b2("Age Category")
```

Graph two Stata commands:

```
. graph twoway connected lar1986 Age || connected lar1940 Age,
    ylab(-4(2)4) xlab(0(10)80) title("Death Rates for 1940 and 1985
    Populations") l2("Log-Age Specific Death Rates") b2("Age
    Category")
```

11. (h) $SMR = (\text{observed deaths}) / (\text{Expected deaths}) = \text{crude} / \text{IA rates}$.
 $SMR(1940) = 1$ (reference) , $SMR(1986) = 1.339 / 1.71 = 0.7821$.
- (i) The 1986 death rates increase faster than the 1940 rate in older age groups.
- (j) Using 1940 rates as the standard:
 Age-adjusted rates: 1940 1.201 per 1000 (= crude rate)
 1986 1.71 per 1000
 (now only 42% greater than 1940 but less than the original 62%)

Note: attached is my Stata log for this problem.

Stata log for problem 10 ps#1, (Pagano #15, p93)

```
use "E:\classes\BC213\data\pagano15.dta", clear
```

```
. * Ok this data I type in by hand  
. list
```

	year	pop	dead	age
1.	1940	10541	494	2
2.	1940	22431	667	9.5
3.	1940	23922	1287	19.5
4.	1940	21339	3696	29.5
5.	1940	18333	11198	39.5
6.	1940	15512	26180	49.5
7.	1940	10572	39071	59.5
8.	1940	6377	44328	69.5
9.	1940	2643	31279	79.5
10.	1986	18152	666	2
11.	1986	33860	1165	9.5
12.	1986	39021	2115	19.5
13.	1986	42779	5604	29.5
14.	1986	33070	14991	39.5
15.	1986	22815	37800	49.5
16.	1986	22232	98805	59.5
17.	1986	17332	146803	69.5
18.	1986	11836	161381	79.5

```
. generate freq=pop/131670
```

```
. replace freq=pop/241097 if year==1986  
(9 real changes made)
```

```
. * to get the age specific rates  
. generate ar=dead/pop
```

```
. list
```

	year	pop	dead	age	freq	ar
1.	1940	10541	494	2	.0800562	.0468646
2.	1940	22431	667	9.5	.1703577	.0297356
3.	1940	23922	1287	19.5	.1816815	.0537998
4.	1940	21339	3696	29.5	.1620643	.173204
5.	1940	18333	11198	39.5	.1392345	.6108111
6.	1940	15512	26180	49.5	.1178097	1.687726
7.	1940	10572	39071	59.5	.0802916	3.695706
8.	1940	6377	44328	69.5	.0484317	6.951231
9.	1940	2643	31279	79.5	.0200729	11.83466
10.	1986	18152	666	2	.0752892	.0366902
11.	1986	33860	1165	9.5	.1404414	.0344064
12.	1986	39021	2115	19.5	.1618477	.0542016
13.	1986	42779	5604	29.5	.1774348	.1309988
14.	1986	33070	14991	39.5	.1371647	.4533111
15.	1986	22815	37800	49.5	.09463	1.656805
16.	1986	22232	98805	59.5	.0922118	4.44427
17.	1986	17332	146803	69.5	.0718881	8.470056
18.	1986	11836	161381	79.5	.0490923	13.63476

Stata log (continued)

```
. * But I cannot calculate adjusted rates easily with data in this form
. * check this out -- I'll reshape the data
```

```
. reshape groups year 1940 1986
```

```
. reshape vars ar freq dead pop
```

```
. reshape cons age
```

```
. reshape wide
```

```
. * You'll need to do this in Stat to get the full effect.
```

```
. list
```

Observation 1

pop1986	18152	dead1986	666	age	2
freq1986	.0752892	ar1986	.0366902	pop1940	10541
dead1940	494	freq1940	.0800562	ar1940	.0468646

Observation 2

pop1986	33860	dead1986	1165	age	9.5
freq1986	.1404414	ar1986	.0344064	pop1940	22431
dead1940	667	freq1940	.1703577	ar1940	.0297356

Observation 3

pop1986	39021	dead1986	2115	age	19.5
freq1986	.1618477	ar1986	.0542016	pop1940	23922
dead1940	1287	freq1940	.1816815	ar1940	.0537998

Observation 4

pop1986	42779	dead1986	5604	age	29.5
freq1986	.1774348	ar1986	.1309988	pop1940	21339
dead1940	3696	freq1940	.1620643	ar1940	.173204

Observation 5

pop1986	33070	dead1986	14991	age	39.5
freq1986	.1371647	ar1986	.4533111	pop1940	18333
dead1940	11198	freq1940	.1392345	ar1940	.6108111

Observation 6

pop1986	22815	dead1986	37800	age	49.5
freq1986	.09463	ar1986	1.656805	pop1940	15512
dead1940	26180	freq1940	.1178097	ar1940	1.687726

Observation 7

pop1986	22232	dead1986	98805	age	59.5
freq1986	.0922118	ar1986	4.44427	pop1940	10572
dead1940	39071	freq1940	.0802916	ar1940	3.695706

Observation 8

pop1986	17332	dead1986	146803	age	69.5
freq1986	.0718881	ar1986	8.470056	pop1940	6377
dead1940	44328	freq1940	.0484317	ar1940	6.951231

Observation 9

pop1986	11836	dead1986	161381	age	79.5
freq1986	.0490923	ar1986	13.63476	pop1940	2643
dead1940	31279	freq1940	.0200729	ar1940	11.83466

. ** Neat huh?

. * ok now the data is in the correct form

. * that is, the populations are side by side like in the book

. * **for the direct adjustment**

. generate DA1986=ar1986*freq1940

. sum DA1986

Variable	Obs	Mean	Std. Dev.	Min	Max
DA1986	9	.1487696	.1635541	.0029373	.4102191

. ** notice Dadj rate=9*.1487696 = 1.3389

. * **similarly for indirect adjustment**

. generate IA1986=ar1940*freq1986

. sum IA1986

Variable	Obs	Mean	Std. Dev.	Min	Max
IA1986	9	.190236	.2267338	.0035284	.5809903

. ** notice Iadj rate=9*.190236 = 1.7121

. ** enjoy !!