

HDS 5230 High performance computing

Homework Week 2

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You will use these datasets to answer some questions listed below. You must be careful to think about what the appropriate denominator is for each question. As you code the answers, be mindful to use the 'high performance' coding approaches in data.table.

```
library(data.table)

## Warning: package 'data.table' was built under R version 3.4.4

data_path = "healthcare2/"
csv_files = list.files(path = data_path, pattern = "*.csv")

readallcsv = function(i){
  assign(
    gsub(".csv", "", csv_files[i]),
    fread(paste0(data_path, csv_files[i])),
    envir = parent.frame()
  )
}

for (i in seq_along(csv_files)) readallcsv(i)
```

- 1) Are men more likely to die than women in this group of patients? Assume people without a date of death in the mortality table are still alive.

Here I recode all patients without a Gender to other.

```
Patient[!Gender %in% c("female", "male"), Gender := "other"]
q1 = Mortality[Patient, on = "PatientID"]
q1[, .(death_percent = sum(!is.na(DateOfDeath))/N), by = Gender][order(-death_percent)]

##      Gender death_percent
## 1:   male      0.3594713
## 2: female      0.3511153
## 3:  other      0.3492670
```

According to the returned data, it seems that males do have a little bit higher chance to die than women in this group of patients, although the difference is nominal.

- 2) I am interested to know if there are patterns in the disease groups across gender. For every patient with at least one outpatient visit, identify if they have been diagnosed with any of the 22 conditions listed in the diseaseMap table at any time point. You will need to consider all three ICD columns in the outpatientVisit file (not just one). Create a table with the rate of disease for each condition for men, women, and all. It should look like this, where the XX% is the percent with the condition:

```
OutpatientVisit = melt(
  OutpatientVisit,
  measure.vars = patterns("^ICD10"),
  id.vars = c("VisitID", "PatientID"),
```

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```

value.name = "ICD10"
)

Npatients = nrow(Patient)

q2 = DiseaseMap[
  OutpatientVisit, on = "ICD10"
][, .N, by = .(PatientID, Condition)][Patient, on = "PatientID"]

q2[
  ,.(condition_N = .N), by = .(Condition, Gender)
][,condition_percent := paste0(condition_N*100/nrow(Patient), "%")][
  order(Condition, Gender)]

```

##	Condition	Gender	condition_N	condition_percent
## 1:	Alcohol	female	737	3.685%
## 2:	Alcohol	male	713	3.565%
## 3:	Alcohol	other	127	0.635%
## 4:	Cancer	female	475	2.375%
## 5:	Cancer	male	446	2.23%
## 6:	Cancer	other	74	0.37%
## 7:	Congestive_heart_failure	female	291	1.455%
## 8:	Congestive_heart_failure	male	501	2.505%
## 9:	Congestive_heart_failure	other	72	0.36%
## 10:	Dementia	female	303	1.515%
## 11:	Dementia	male	266	1.33%
## 12:	Dementia	other	48	0.24%
## 13:	Depression	female	1182	5.91%
## 14:	Depression	male	751	3.755%
## 15:	Depression	other	173	0.865%
## 16:	Diabetes_with_complications	female	397	1.985%
## 17:	Diabetes_with_complications	male	344	1.72%
## 18:	Diabetes_with_complications	other	70	0.35%
## 19:	Diabetes_without_complications	female	974	4.87%
## 20:	Diabetes_without_complications	male	875	4.375%
## 21:	Diabetes_without_complications	other	146	0.73%
## 22:	Drugs	female	387	1.935%
## 23:	Drugs	male	343	1.715%
## 24:	Drugs	other	59	0.295%
## 25:	HIV	female	54	0.27%
## 26:	HIV	male	56	0.28%
## 27:	HIV	other	15	0.075%
## 28:	Hypertension	female	2712	13.56%
## 29:	Hypertension	male	2866	14.33%
## 30:	Hypertension	other	441	2.205%
## 31:	LiverMild	female	90	0.45%
## 32:	LiverMild	male	83	0.415%
## 33:	LiverMild	other	11	0.055%
## 34:	LiverSevere	female	460	2.3%
## 35:	LiverSevere	male	466	2.33%
## 36:	LiverSevere	other	87	0.435%
## 37:	Metastatic_solid_tumour	female	312	1.56%
## 38:	Metastatic_solid_tumour	male	309	1.545%
## 39:	Metastatic_solid_tumour	other	38	0.19%

## 40:	Myocardial_infarction	female	300	1.5%
## 41:	Myocardial_infarction	male	526	2.63%
## 42:	Myocardial_infarction	other	74	0.37%
## 43:	Obesity	female	1745	8.725%
## 44:	Obesity	male	1247	6.235%
## 45:	Obesity	other	250	1.25%
## 46:	Paralysis	female	139	0.695%
## 47:	Paralysis	male	104	0.52%
## 48:	Paralysis	other	25	0.125%
## 49:	Peptic_ulcer_disease	female	97	0.485%
## 50:	Peptic_ulcer_disease	male	80	0.4%
## 51:	Peptic_ulcer_disease	other	14	0.07%
## 52:	Peripheral_vascular_disease	female	229	1.145%
## 53:	Peripheral_vascular_disease	male	200	1%
## 54:	Peripheral_vascular_disease	other	47	0.235%
## 55:	Pulmonary	female	672	3.36%
## 56:	Pulmonary	male	647	3.235%
## 57:	Pulmonary	other	114	0.57%
## 58:	Renal	female	345	1.725%
## 59:	Renal	male	305	1.525%
## 60:	Renal	other	48	0.24%
## 61:	Rheumatic	female	124	0.62%
## 62:	Rheumatic	male	98	0.49%
## 63:	Rheumatic	other	21	0.105%
## 64:	Stroke	female	253	1.265%
## 65:	Stroke	male	271	1.355%
## 66:	Stroke	other	39	0.195%
## 67:	<NA>	female	8229	41.145%
## 68:	<NA>	male	7743	38.715%
## 69:	<NA>	other	1370	6.85%
##	Condition	Gender	condition_N	condition_percent

I assume the denominator here is the number of patients : 20000

- 3) Calculate the mortality rate for every year between 2005 and 2018. Is it generally increasing, or decreasing? Assume patients are only at risk of death as of their first visit (in the outpatient Visit file). Once they have died, they are no longer at risk in subsequent years

```
q3 = Mortality[Patient, on = "PatientID"]
q3[, year := as.integer(substr(DateOfDeath, 1, 4))]
q3 = q3[, .(N_death = .N), by = year][order(year)][!is.na(year)]

q3[, cum_death := cumsum(N_death)
  ][, atrisk := 20000 - shift(cum_death, fill = 0, type = "lag")
  ][, mortality_rate := N_death*100/atrisk]

q3
```

##	year	N_death	cum_death	atrisk	mortality_rate
## 1:	2005	79	79	20000	0.395000
## 2:	2006	235	314	19921	1.179660
## 3:	2007	356	670	19686	1.808392
## 4:	2008	423	1093	19330	2.188308
## 5:	2009	479	1572	18907	2.533453
## 6:	2010	567	2139	18428	3.076840
## 7:	2011	605	2744	17861	3.387268

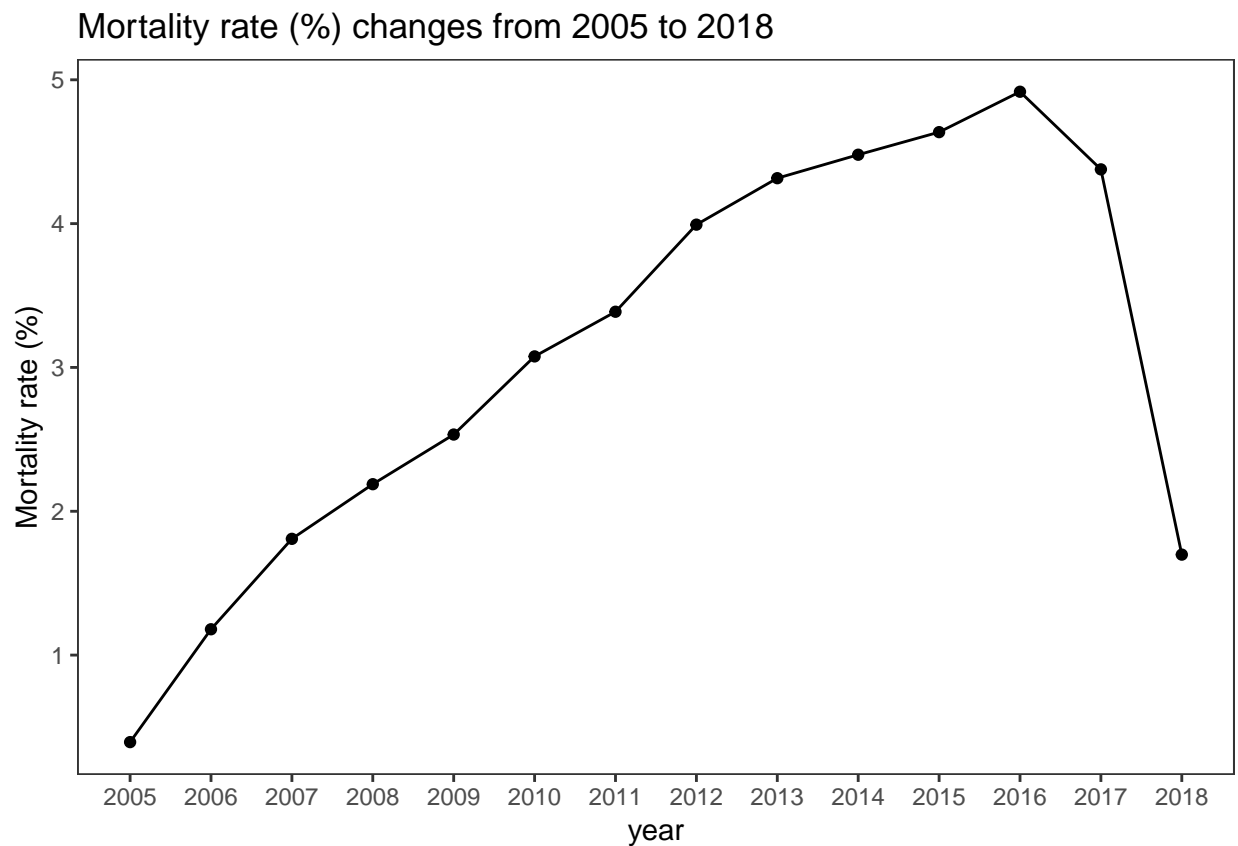
```
## 8: 2012      689      3433 17256      3.992814
## 9: 2013      715      4148 16567      4.315809
## 10: 2014     710      4858 15852      4.478930
## 11: 2015     702     5560 15142      4.636111
## 12: 2016     710     6270 14440      4.916898
## 13: 2017     601     6871 13730      4.377276
## 14: 2018     223     7094 13129      1.698530
```

```
require(ggplot2)
```

```
## Loading required package: ggplot2
```

```
## Warning: package 'ggplot2' was built under R version 3.4.4
```

```
ggplot(q3, aes(year, mortality_rate)) +
  geom_point() + geom_line() +
  scale_x_continuous("year", labels = 2005:2018, breaks = 2005:2018) +
  labs(title = "Mortality rate (%) changes from 2005 to 2018") +
  ylab("Mortality rate (%)") + theme_test()
```



According to the time trend plot, the mortality rate has been generally increasing, while it experienced a major drop in the recent two years (2017 and 2018).

- a. This is a harder question to answer than at first glance. What should the denominator of patients be for every year? How will you calculate it?

From my understanding, the denominator should be the patients at risk in the specific year (who were still alive). I calculated it by excluding the patients till the last year.