

# 1.

```
library(tidyverse)
```

```
## —— Attaching core tidyverse packages —— tidyverse 2.0.0 ——
## ✓ dplyr      1.1.4      ✓ readr      2.1.5
## ✓ forcats    1.0.0      ✓ stringr    1.5.1
## ✓ ggplot2    3.5.1      ✓ tibble     3.2.1
## ✓ lubridate  1.9.3      ✓ tidyr      1.3.1
## ✓ purrr      1.0.2
## —— Conflicts ——
——— tidyverse_conflicts() ——
## ✗ dplyr::filter() masks stats::filter()
## ✗ dplyr::lag()     masks stats::lag()
## ⓘ Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
```

```
ckm_nodes <- read_csv('data/ckm_nodes.csv')
```

```
## Rows: 246 Columns: 13
## —— Column specification ——
—————
## Delimiter: ","
## chr (10): city, medical_school, attend_meetings, free_time_with, discuss_med...
## dbl (3): adoption_date, medical_journals, drs_among_three_best_friends
##
## ⓘ Use `spec()` to retrieve the full column specification for this data.
## ⓘ Specify the column types or set `show_col_types = FALSE` to quiet this message.
```

```
noinfor <- which(is.na(ckm_nodes$adoption_date))
ckm_nodes <- ckm_nodes[-noinfor, ]
ckm_network <- read.table('data/ckm_network.dat')
ckm_network <- ckm_network[-noinfor,-noinfor]
```

# 2.

ckm\_nodes 中记录了125个医生在17个月中的数据，因此行数应为  $17 * 125 = 2150$ 。需要记录的信息有4项，再加上医生序号及月份，故共需要6列。

```

doc_mon$began_prescribing <-
  ckm_nodes$adoption_date[doc_mon$doctor] == doc_mon$month
doc_mon$began_prescribing_before <-
  ckm_nodes$adoption_date[doc_mon$doctor] < doc_mon$month

tb <- t(ckm_nodes$adoption_date)[rep(1, nrow(doc_mon)), ]
doc_mon$n_began_prescribing_before <-
  rowSums(ckm_network[doc_mon$doctor, ] & (tb < doc_mon$month))
doc_mon$n_began_prescribing_or_before <-
  rowSums(ckm_network[doc_mon$doctor, ] & (tb <= doc_mon$month))

head(doc_mon)

```

```

##   doctor month began_prescribing began_prescribing_before
## 1      1     1             TRUE             FALSE
## 2      1     2             FALSE             TRUE
## 3      1     3             FALSE             TRUE
## 4      1     4             FALSE             TRUE
## 5      1     5             FALSE             TRUE
## 6      1     6             FALSE             TRUE
##   n_began_prescribing_before n_began_prescribing_or_before
## 1                        0                        1
## 2                        1                        1
## 3                        1                        2
## 4                        2                        3
## 5                        3                        3
## 6                        3                        3

```

### 3.

#### a.

```
max(rowSums(ckm_network))
```

```
## [1] 20
```

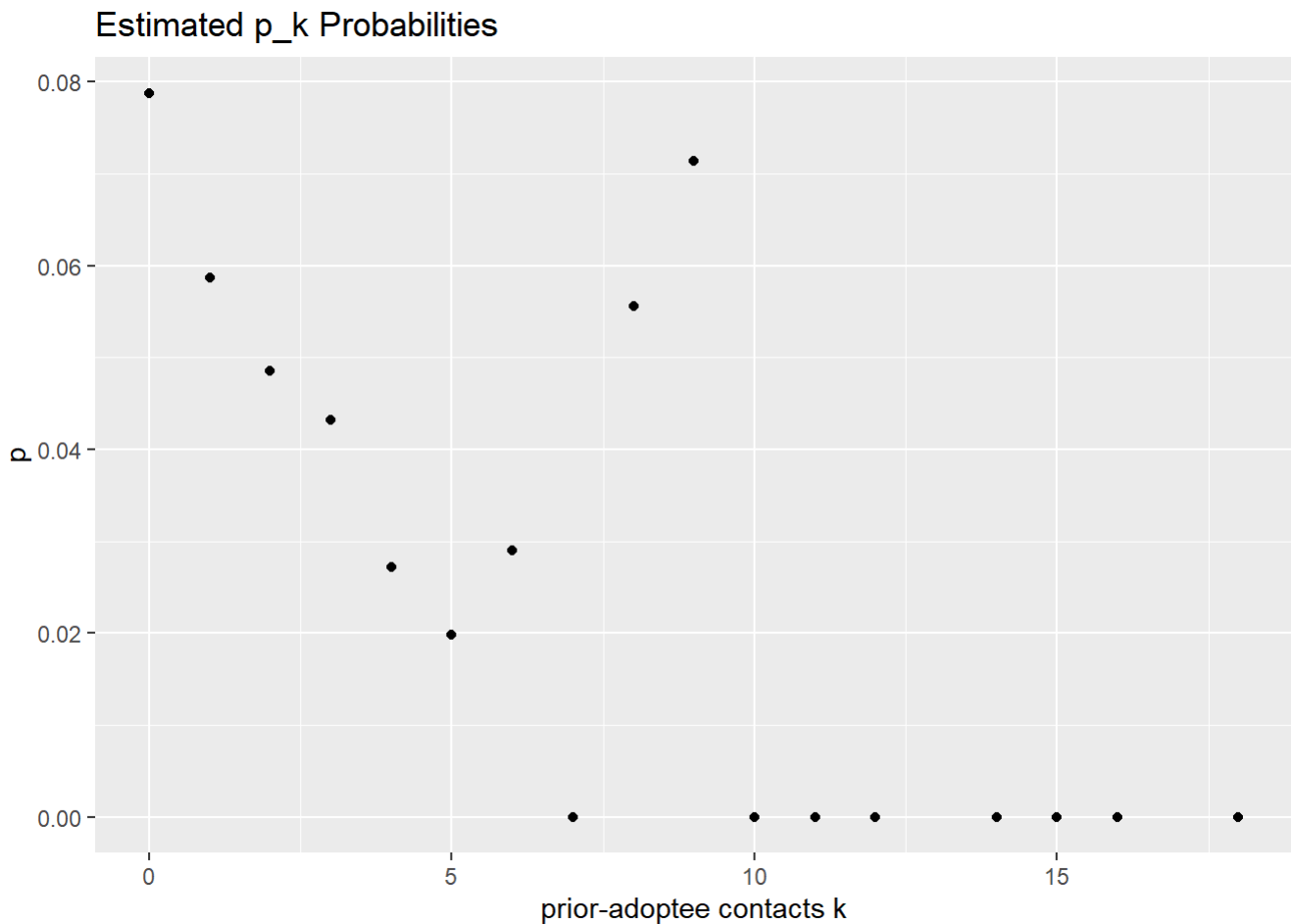
数据中一个医生最多有 20 个联系人，因此 k 只能取 0 ~ 20 共21个量。

```

  rename(k = n_began_prescribing_before)

ggplot(data = p_data) +
  geom_point(aes(x = k, y = p_k)) +
  labs(title = "Estimated p_k Probabilities",
        x = "prior-adoptee contacts k",
        y = "p")

```



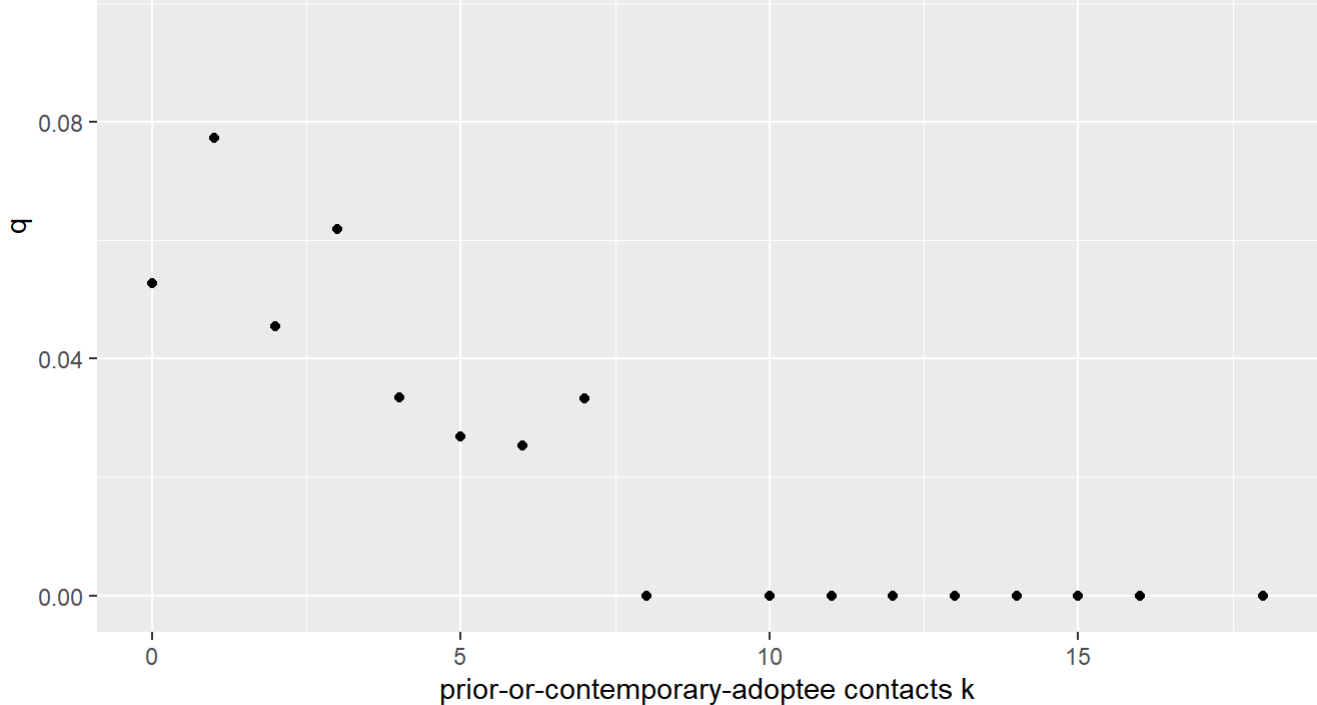
**C.**

```

q_data <- doc_mon |>
  group_by(n_began_prescribing_or_before) |>
  summarise(num = sum(began_prescribing), total = n()) |>
  mutate(q_k = num / total) |>
  rename(k = n_began_prescribing_or_before)

ggplot(data = q_data) +
  geom_point(aes(x = k, y = q_k)) +
  labs(title = "Estimated q_k Probabilities",
        x = "prior-or-contemporary-adoptee contacts k",
        y = "q")

```



## 4. ### a.

```
a.model <- lm(p_k ~ k, data = p_data)
a.model$coefficients
```

```
## (Intercept)          k
## 0.056932428 -0.003799739
```

b.

若  $b > 0$ ，则起初时随着  $k$  的增大， $p$  的增长速率会逐渐上升，当  $k$  增大到一定程度时， $p$  的增长速率放缓，最终停止增长。

```
b.func <- function(para, X) {
  return(1 - 1 / (1 + exp(para[1] + para[2] * X)))
}

b.model <- nls(p_k ~ b.func(para, k),
  data = p_data,
  start = list(para = c(0, 0)))
summary(b.model)
```

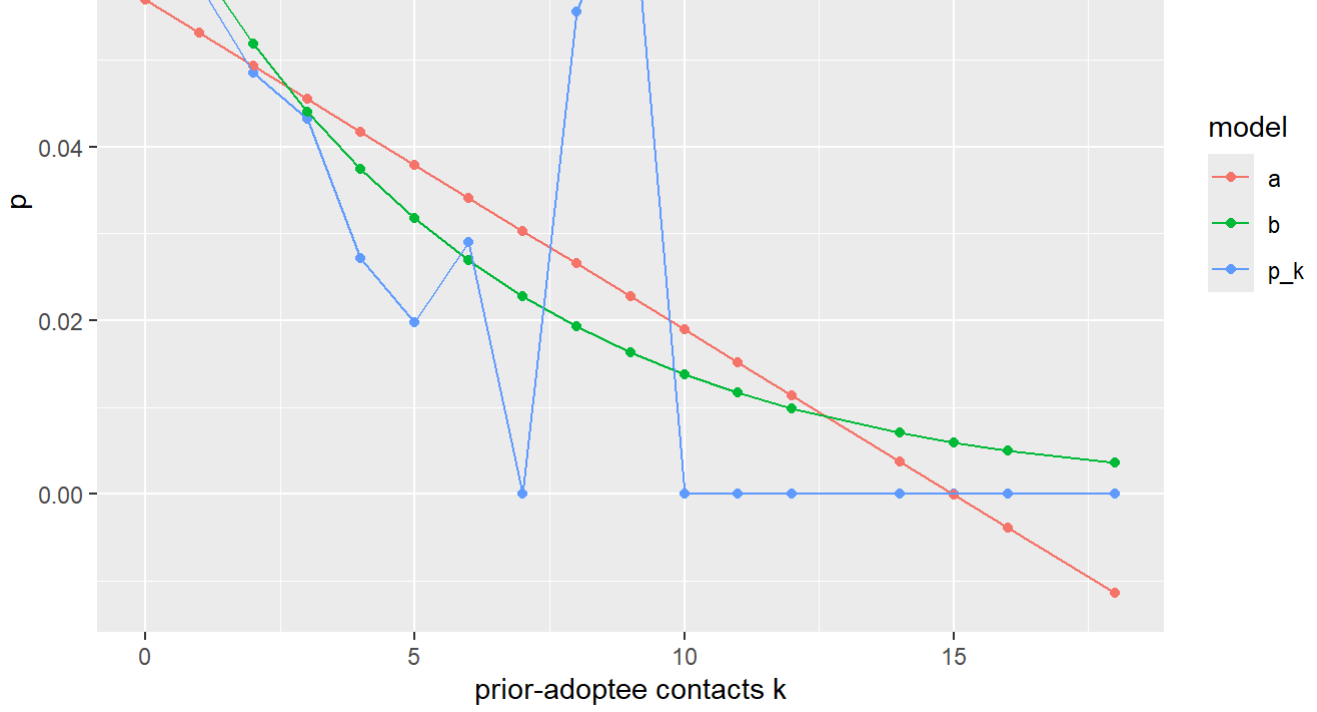
```
## para2 -0.17051    0.05371   -3.174   0.00628 **  
## ---  
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1  
##  
## Residual standard error: 0.01957 on 15 degrees of freedom  
##  
## Number of iterations to convergence: 7  
## Achieved convergence tolerance: 1.053e-07
```

## C.

```
p_data <- p_data |>  
  mutate(a = predict(a.model, p_data), b = predict(b.model, p_data))  
  
p_tidy <- p_data |>  
  select(-num, -total) |>  
  gather(model, p, -k)
```

```
## Warning: attributes are not identical across measure variables; they will be  
## dropped
```

```
p_tidy |>  
  ggplot(aes(x = k, y = p, col = model)) +  
  geom_point() +  
  geom_line() +  
  labs(x = "prior-adoptee contacts k")
```



4(b) 中的模型似乎更加符合计算出的  $p_k$ 。模型 a 中  $p$  的取值为  $[-\infty, \infty]$ ，而模型 b 中  $p$  的取值为  $[0, 1]$ ，更加贴合实际。