Loess, Polynomial Regression and Spline.rmd

Loess, Polynomial Regression and Spline

Goal1. Loess 모델 생성

data.test <- DATA[2001:3000,]</pre>

```
Goal2. Polynomial Regression 모델 생성
Goal3. Spline 모델 생성
Goal4. 각 모델의 비교
Data Loading and Setting
데이터 로딩
DATA <- read.table("KMOOC_2_03_dataset_salary.txt", header=TRUE)
데이터 탐색
DATA[1:10,]
     age
          salary
## 1 18 75.04315
## 2 24 70.47602
## 3 45 130.98218
## 4 43 154.68529
## 5 50 75.04315
## 6 54 127.11574
## 7 44 169.52854
     30 111.72085
## 9 41 118.88436
## 10 52 128.68049
dim(DATA)
## [1] 3000
              2
데이터 분리
data.train <- DATA[1:2000,]</pre>
```

Loess

Loess 모델 생성

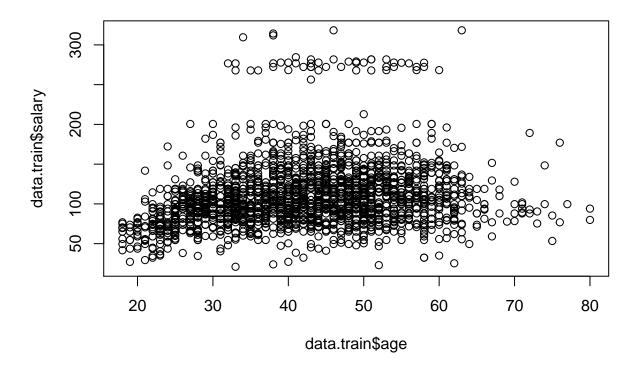
```
f.ls <- loess(salary ~ age, data=data.train)
f.ls

## Call:
## loess(formula = salary ~ age, data = data.train)
##

## Number of Observations: 2000
## Equivalent Number of Parameters: 4.98
## Residual Standard Error: 39.23

age와 salary의 산점도 시각화

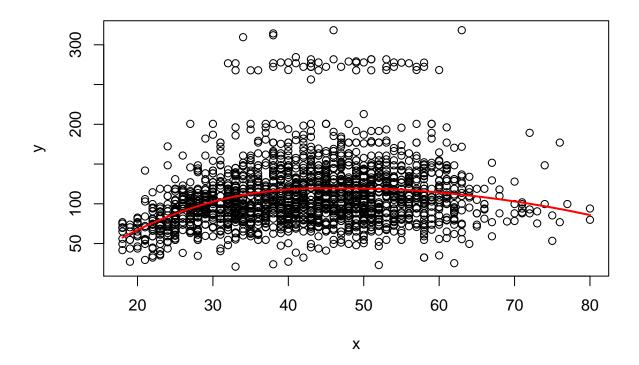
plot(data.train$age, data.train$salary)
```



plot_fitted 함수 정의

```
plot_fitted <- function(x, y, yhat) {
  plot(x, y)
  o <- order(x)
  lines(x[o], yhat[o], type='l', col=2, lwd=2)
}</pre>
```

```
plot_fitted(data.train$age, data.train$salary, f.ls$fitted)
```



Loess 모델의 RMSE 계산

```
y.train.ls <- predict(f.ls, newdata=data.train)
rmse.train.ls <- sqrt(mean((y.train.ls - data.train$salary)^2))
rmse.train.ls

## [1] 39.17384

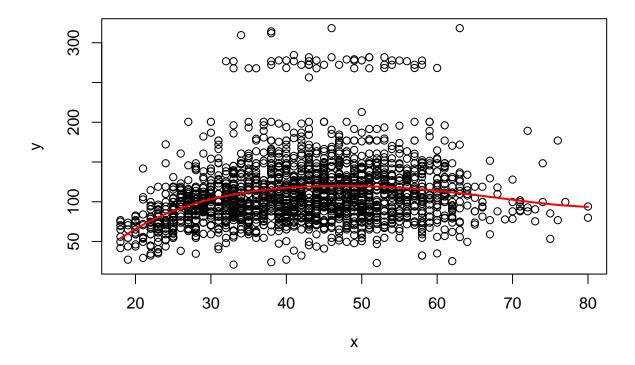
y.test.ls <- predict(f.ls, newdata=data.test)
rmse.test.ls <- sqrt(mean((y.test.ls - data.test$salary)^2))
rmse.test.ls</pre>
```

[1] 41.28488

Polynomial Regression

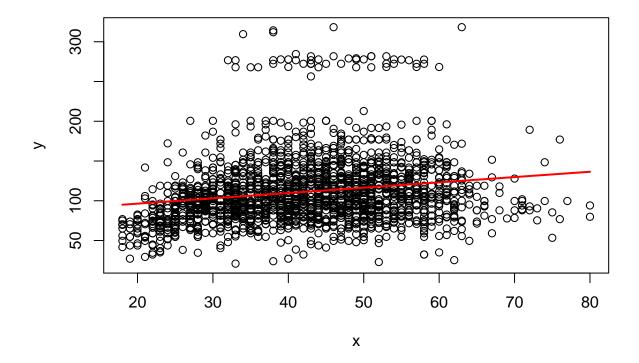
Polynomial Regression 모델 생성

```
f.pl <- lm(salary ~ poly(age, 5, raw=TRUE), data=data.train)</pre>
summary(f.pl)
##
## Call:
## lm(formula = salary ~ poly(age, 5, raw = TRUE), data = data.train)
## Residuals:
      Min
                1Q Median
                                3Q
                                       Max
## -96.171 -24.058 -5.509 14.845 207.607
##
## Coefficients:
##
                               Estimate Std. Error t value Pr(>|t|)
                             -1.862e+02 1.980e+02 -0.941
## (Intercept)
                                                              0.347
## poly(age, 5, raw = TRUE)1 2.311e+01 2.474e+01 0.934
                                                              0.350
## poly(age, 5, raw = TRUE)2 -7.368e-01 1.184e+00 -0.622 0.534
## poly(age, 5, raw = TRUE)3 1.277e-02 2.717e-02 0.470
                                                              0.638
## poly(age, 5, raw = TRUE)4 -1.197e-04 2.999e-04 -0.399
                                                              0.690
## poly(age, 5, raw = TRUE)5 4.608e-07 1.277e-06 0.361
                                                              0.718
##
## Residual standard error: 39.23 on 1994 degrees of freedom
## Multiple R-squared: 0.08786, Adjusted R-squared: 0.08558
## F-statistic: 38.42 on 5 and 1994 DF, p-value: < 2.2e-16
Polynomial Regression 모델의 RMSE 계산
y.train.pl <- predict(f.pl, newdata=data.train)</pre>
rmse.train.pl <- sqrt(mean((y.train.pl - data.train$salary)^2))</pre>
rmse.train.pl
## [1] 39.1756
y.test.pl <- predict(f.pl, newdata=data.test)</pre>
rmse.test.pl <- sqrt(mean((y.test.pl - data.test$salary)^2))</pre>
rmse.test.pl
## [1] 41.3284
Polynomial Regression 모델 시각화
plot_fitted(data.train$age, data.train$salary, f.pl$fitted)
```



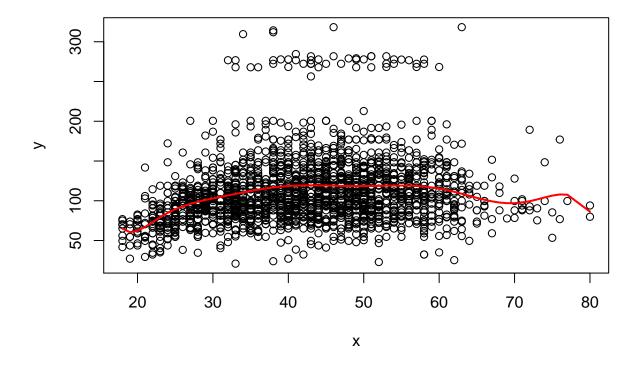
1차항 Polynomial Regression 모델 시각화

```
f.pl.1 <- lm(salary ~ poly(age, 1, raw=TRUE), data=data.train)
plot_fitted(data.train$age, data.train$salary, f.pl.1$fitted)</pre>
```



10차항 Polynomial Regression 모델 시각화

```
f.pl.10 <- lm(salary ~ poly(age, 10, raw=TRUE), data=data.train)
plot_fitted(data.train$age, data.train$salary, f.pl.10$fitted)</pre>
```



Spline

splines 패키지 attach

```
library(splines)
```

cut off point 생성

```
quantile(data.train$age, c(0, 0.25, 0.5, 0.75, 1))

## 0% 25% 50% 75% 100%

## 18 33 42 51 80

cutpt <- quantile(data.train$age, prob=seq(0, 1, by=0.25))
cutpt <- cutpt[2:(length(cutpt)-1)]
cutpt</pre>
```

```
## 25% 50% 75%
## 33 42 51
```

Spline 모델 생성

```
f.sp <- lm(salary ~ bs(age, knots=cutpt), data=data.train)</pre>
```

Spline 모델의 RMSE 계산

```
y.train.sp <- predict(f.sp, newdata=data.train)
rmse.train.sp <- sqrt(mean((y.train.sp - data.train$salary)^2))
rmse.train.sp</pre>
```

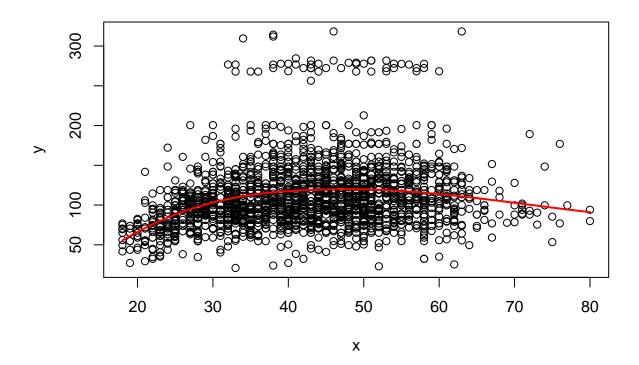
[1] 39.1753

```
y.test.sp <- predict(f.sp, newdata=data.test)
rmse.test.sp <- sqrt(mean((y.test.sp - data.test$salary)^2))
rmse.test.sp</pre>
```

[1] 41.32029

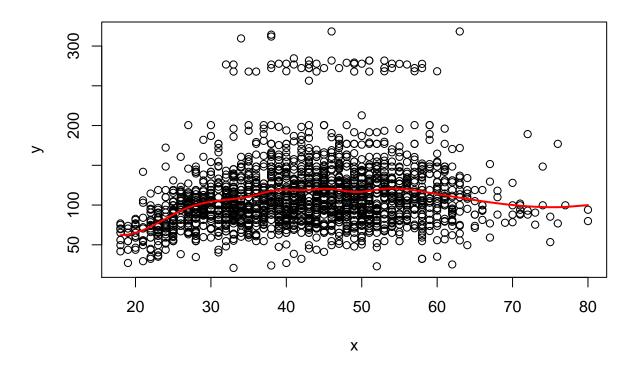
Spline 모델 시각화

```
plot_fitted(data.train$age, data.train$salary, f.sp$fitted)
```



10개의 cutpt를 갖는 Spline 모델 시각화

```
cutpt <- quantile(data.train$age, prob=seq(0, 1, by=0.1))
cutpt <- cutpt[2:(length(cutpt)-1)]
f.sp.10 <- lm(salary ~ bs(age, knots=cutpt), data=data.train)
plot_fitted(data.train$age, data.train$salary, f.sp.10$fitted)</pre>
```



각 모델 비교

각 모델의 RMSE 비교 및 시각화

