

## Part A

```
> lm_fit=lm(newSiribFperc~Weight+Height+NeckC+ChestC+AbdomenC+HipC+ThighC+KneeC+AnkleC+BicepsC+ForearmC+WristC+dummy)
> summary(lm_fit)
```

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
--	----------	------------	---------	----------

dummy	0.10052	0.69435	0.145	0.88502
-------	---------	---------	-------	---------

```
> lm_1=update(lm_full, .~.-Over45)
```

```
> summary(lm_1)
```

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
--	----------	------------	---------	----------

KneeC	0.07088	0.24386	0.291	0.77157
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---

```
> lm_2=update(lm_1, .~.-KneeC)
```

```
> summary(lm_2)
```

Coefficients:

ThighC	0.08890	0.13818	0.643	0.52063
--------	---------	---------	-------	---------

---

```
> lm_3=update(lm_2, .~.-ThighC)
```

```
> summary(lm_3)
```

Coefficients:

Weight	-0.03504	0.06383	-0.549	0.583485
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```
> lm_4=update(lm_3, .~.-Weight)
```

```
> summary(lm_4)
```

Coefficients:

ForearmC	0.20957	0.20698	1.013	0.312311
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```
> lm_5=update(lm_4, .~.-ForearmC)
```

```
> summary(lm_5)
```

Coefficients:

AnkleC	0.23520	0.21275	1.106	0.27003
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```
> lm_6=update(lm_5, .~.-AnkleC)
```

```
> summary(lm_6)
```

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
--	----------	------------	---------	----------

(Intercept)	20.96118	8.23097	2.547	0.01150 *
Height	-0.37610	0.12712	-2.959	0.00340 **
NeckC	-0.36516	0.22459	-1.626	0.10527
ChestC	-0.18595	0.09065	-2.051	0.04130 *
AbdomenC	0.99730	0.07767	12.840	< 2e-16 ***
HipC	-0.19133	0.10123	-1.890	0.05995 .
BicepsC	0.33298	0.15265	2.181	0.03012 *
WristC	-1.46315	0.46914	-3.119	0.00204 **

---

Answer: From The each step to regression, we can compare the largest Pvalue to delete from the model, the order is **dummy(Over45), Kneec, ThighC, weight, ForearmC, Anklec**

The regression model is  $\text{newSiriBFperc} = 20.96118 - 0.37610\text{Height} - 0.36516\text{NeckC} - 0.18595\text{ChestC} + 0.99730\text{AbdomenC} - 0.19133\text{HipC} + 0.33298\text{BicepsC} - 1.46315\text{WristC}$

Part B

```
> lm_empty=lm(Bodyfate$newSiriBFperc~1)
> summary(lm_empty)
```

Residuals:

Min	1Q	Median	3Q	Max
-18.9859	-6.5420	0.2306	6.2217	28.5015

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	18.9859	0.5286	35.92	<2e-16 ***

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Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 8.357 on 249 degrees of freedom

```
> pvalue_for1=numeric(13)
> names(pvalue_for1)=c("weight","Height","NeckC","ChestC","AbdomenC","HipC","ThighC","KneeC","AnkleC","BicepsC","ForearmC","WristC","Over45")
weight      Height      NeckC      ChestC      AbdomenC      HipC
1.105322e-25 4.464991e-01 3.452486e-15 1.834532e-35 1.027911e-59 1.198462e-27
ThighC      KneeC      AnkleC      BicepsC      ForearmC      WristC
2.473884e-20 1.536127e-15 1.106314e-04 1.910743e-15 9.617845e-09 3.388341e-07
Over45
3.279761e-02
```

```
> lm_for1=update(lm_empty,~.+AbdomenC)
> summary(lm_for1)
```

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	-42.3067	2.8230	-14.99	<2e-16 ***
AbdomenC	0.6642	0.0304	21.84	<2e-16 ***

```
> pvalue_for2=numeric(12)
> names(pvalue_for2)=c("weight","Height","NeckC","ChestC","HipC","ThighC","KneeC","AnkleC","BicepsC","ForearmC","WristC","Over45")
weight      Height      NeckC      ChestC      HipC      ThighC
2.354923e-10 1.730149e-08 1.407977e-06 1.857095e-04 1.439466e-05 2.112094e-02
KneeC      AnkleC      BicepsC      ForearmC      WristC      Over45
1.022381e-04 8.833134e-03 3.876392e-02 1.398205e-02 1.697697e-09 4.905053e-01
> lm_for2=update(lm_for1,~.+weight)
> summary(lm_for2)
```

Call:

Coefficients:

```
      Estimate Std. Error t value Pr(>|t|)
(Intercept) -47.41951    2.71976  -17.43 < 2e-16 ***
AbdomenC      0.99759    0.05774   17.28 < 2e-16 ***
weight       -0.14408    0.02180   -6.61 2.35e-10 ***
---
```

```
> pvalue_for3=numeric(11)
```

```
> names(pvalue_for3)=c("Height","NeckC","ChestC","HipC","ThighC","Kneec","AnkleC",
"BicepsC","ForearmC","WristC","Over45")
```

```
      Height      NeckC      ChestC      HipC      ThighC      Kneec
0.1094944470 0.0495148635 0.3343920729 0.7687362450 0.0282540883 0.8630623544
      AnkleC      BicepsC      ForearmC      WristC      Over45
0.5455442304 0.0529244274 0.3015792059 0.0005940908 0.0827633997
> lm_for3=update(lm_for2,~.+WristC)
> summary(lm_for3)
```

Coefficients:

```
      Estimate Std. Error t value Pr(>|t|)
(Intercept) -25.23010    6.90997  -3.651 0.000319 ***
AbdomenC      0.97919    0.05673   17.261 < 2e-16 ***
weight       -0.09985    0.02483   -4.022 7.69e-05 ***
WristC       -1.55701    0.44748   -3.479 0.000594 ***
---
```

```
> pvalue_for4=numeric(10)
```

```
> names(pvalue_for4)=c("Height","NeckC","ChestC","HipC","ThighC","Kneec","AnkleC",
"BicepsC","ForearmC","Over45")
```

```
      Height      NeckC      ChestC      HipC      ThighC      Kneec      AnkleC
0.08753103 0.43136745 0.44007787 0.73179176 0.12301391 0.56900979 0.17407271
      BicepsC      ForearmC      Over45
0.02001777 0.07908759 0.59822035
> lm_for4=update(lm_for3,~.+BicepsC)
> summary(lm_for4)
```

Call:

```
lm(formula = Bodyfate$newSirIBFperc ~ AbdomenC + weight + WristC +
    BicepsC)
```

```
      Estimate Std. Error t value Pr(>|t|)
(Intercept) -30.45503    7.20232  -4.229 3.32e-05 ***
AbdomenC      0.99108    0.05645   17.558 < 2e-16 ***
weight       -0.13214    0.02821   -4.685 4.65e-06 ***
WristC       -1.65671    0.44550   -3.719 0.000248 ***
BicepsC       0.36298    0.15503    2.341 0.020018 *
```

```
> pvalue_for5=numeric(9)
```

```
> names(pvalue_for5)=c("Height","NeckC","ChestC","HipC","ThighC","Kneec","AnkleC",
"ForearmC","Over45")
```

```
      Height      NeckC      ChestC      HipC      ThighC      Kneec
0.26942476 -0.28506613 0.30902147 0.73958012 0.32316175 0.49763072
      AnkleC      ForearmC      Over45
```

0.12863924 0.30331958 0.02168526

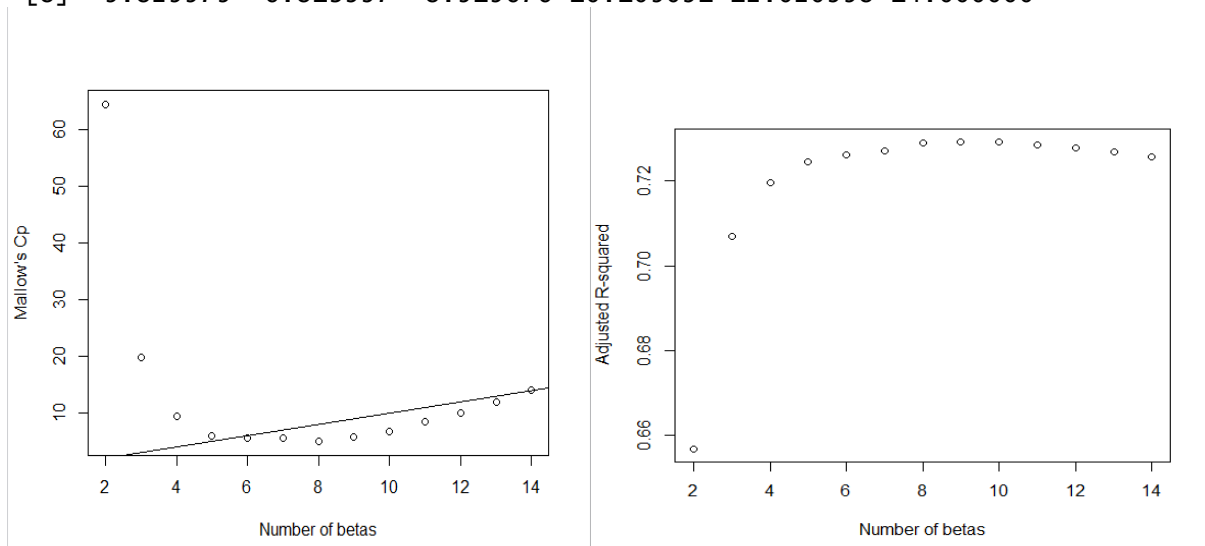
We first write the empty model, and computer each Pvalue of variable, we choose the least Pvalue variable to the empty model in each step, the order is **AbdomenC ,weight ,wristC,BicepsC**  
 The regression model is **newSiriBFperc =0.99108AbdomenC -0.13214weight -1.65671 wristC+0.36298BicepsC-30.45503**

Partc

```
> subset=regsubsets(newSiriBFperc~Weight+Height+NeckC+ChestC+AbdomenC+HipC+ThighC+KneeC+AnkleC+BicepsC+ForearmC+WristC+Over45,method="exhaustive",nbest=1,nvmax=13,data=Bodyfate)
> sum_subset$which
```

	(Intercept)	weight	Height	NeckC	ChestC	AbdomenC	HipC	ThighC	KneeC	AnkleC	BicepsC	ForearmC	wristC	Over45
1	TRUE	FALSE	FALSE	FALSE	FALSE	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE
2	TRUE	TRUE	FALSE	FALSE	FALSE	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE
3	TRUE	TRUE	FALSE	FALSE	FALSE	TRUE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	TRUE	FALSE
4	TRUE	TRUE	FALSE	FALSE	FALSE	TRUE	FALSE	FALSE	FALSE	FALSE	TRUE	FALSE	TRUE	FALSE
5	TRUE	TRUE	FALSE	FALSE	FALSE	TRUE	FALSE	FALSE	FALSE	TRUE	TRUE	FALSE	TRUE	FALSE
6	TRUE	FALSE	TRUE	FALSE	TRUE	TRUE	TRUE	FALSE	FALSE	FALSE	TRUE	FALSE	TRUE	FALSE
7	TRUE	FALSE	TRUE	TRUE	TRUE	TRUE	TRUE	FALSE	FALSE	FALSE	TRUE	FALSE	TRUE	FALSE
8	TRUE	FALSE	TRUE	TRUE	TRUE	TRUE	TRUE	FALSE	FALSE	TRUE	TRUE	FALSE	TRUE	FALSE
9	TRUE	FALSE	TRUE	TRUE	TRUE	TRUE	TRUE	FALSE	FALSE	TRUE	TRUE	TRUE	TRUE	FALSE
10	TRUE	TRUE	TRUE	TRUE	TRUE	TRUE	TRUE	FALSE	FALSE	TRUE	TRUE	TRUE	TRUE	FALSE
11	TRUE	TRUE	TRUE	TRUE	TRUE	TRUE	TRUE	TRUE	FALSE	TRUE	TRUE	TRUE	TRUE	FALSE
12	TRUE	TRUE	TRUE	TRUE	TRUE	TRUE	TRUE	TRUE	TRUE	TRUE	TRUE	TRUE	TRUE	FALSE
13	TRUE	TRUE	TRUE	TRUE	TRUE	TRUE	TRUE	TRUE	TRUE	TRUE	TRUE	TRUE	TRUE	TRUE

```
> R2_adj
[1] 0.6566308 0.7070561 0.7196620 0.7246780 0.7261585 0.7272627 0.7290949
[8] 0.7293434 0.7293717 0.7285817 0.7279145 0.7268638 0.7257308
> C_p
[1] 64.481671 19.818005 9.443323 5.940477 5.619517 5.642793 5.031670
[8] 5.825575 6.813997 8.515676 10.105091 12.020958 14.000000
```

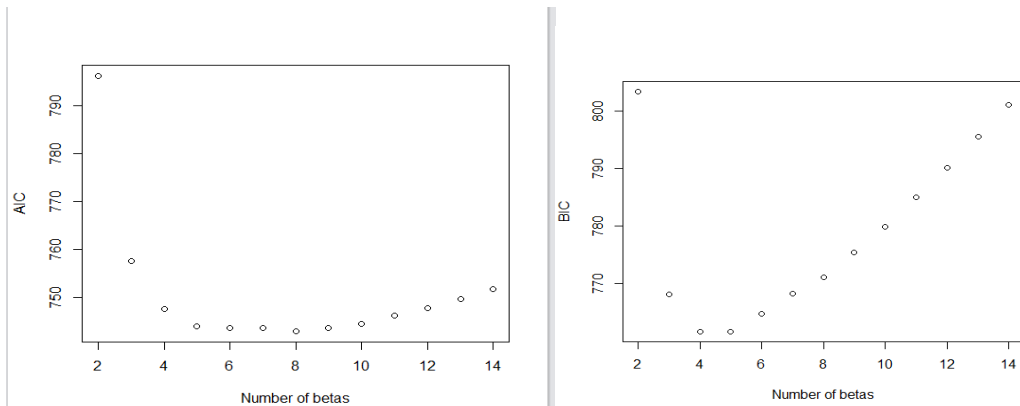


```
> aic_p
```

```
[1] 796.3263 757.6099 747.5994 744.0674 743.6969 743.6602 742.9440 743.6794
[9] 744.6138 746.2987 747.8643 749.7752 751.7530
```

```
> bic_p
```

```
[1] 803.3692 768.1743 761.6653 761.6747 764.8257 768.3104 771.1157 775.3726
[9] 779.8284 785.0347 790.1218 795.5542 801.0534
```



For AIC, we choose model 7 (the AIC IS LEAST) 7 predictors are: Height NeckC ChestC AbdomenC HipC ForearmC WristC.

For BIC, we choose model 3 (the BIC is the least value). There are 4 predictors: weight AbdomenC wristC

For CP, we choose model 5, because the variable is least and approach 5. There are 5 predictors: weight AbdomenC AnkleC BicepsC wristC

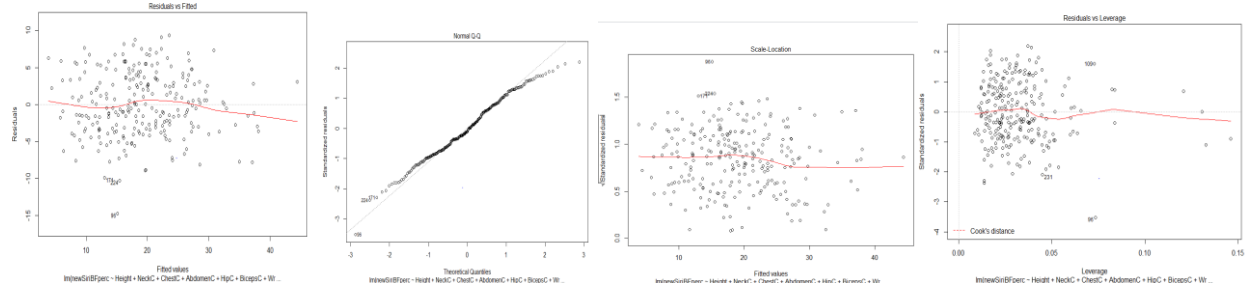
For adjust  $R^2$ , we choose model 9, the value of  $R^2$  adjust is the highest. There are 9 predictors: Height NeckC ChestC AbdomenC HipC ThighC AnkleC BicepsC ForearmC wristC

For AIC chosen model, the  $R^2$  adjusted is very close to adjust  $R^2$  chosen model, but the predictor is small than that, it much better.

For the cp of BIC model also close to  $P=4$ , (it is less close compared to CP best model). But the predictors is less, might be suitable for simplistics.

## Part D

### For Amodel

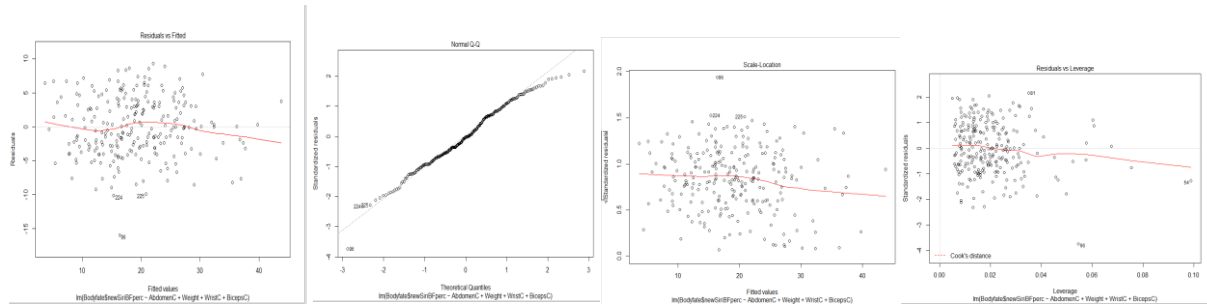


shapiro-wilk normality test

```
data: lm.backward$residuals
```

w = 0.99289, p-value = 0.2775

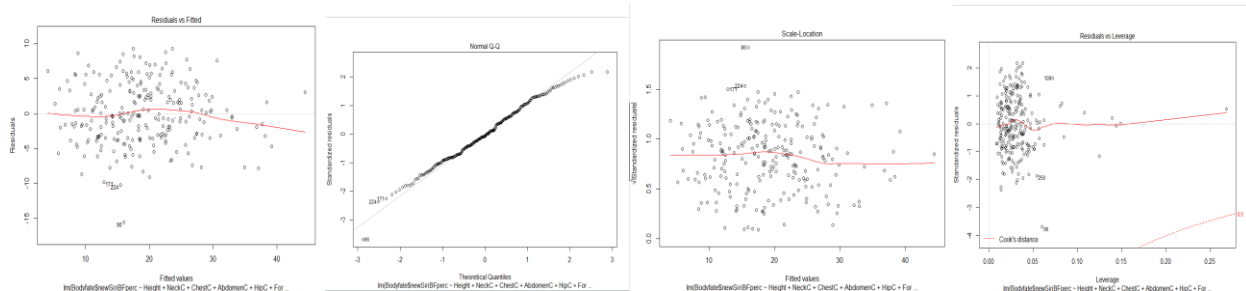
For model B



Shapiro-wilk normality test

data: lm.forward\$residuals  
w = 0.98869, p-value = 0.04722

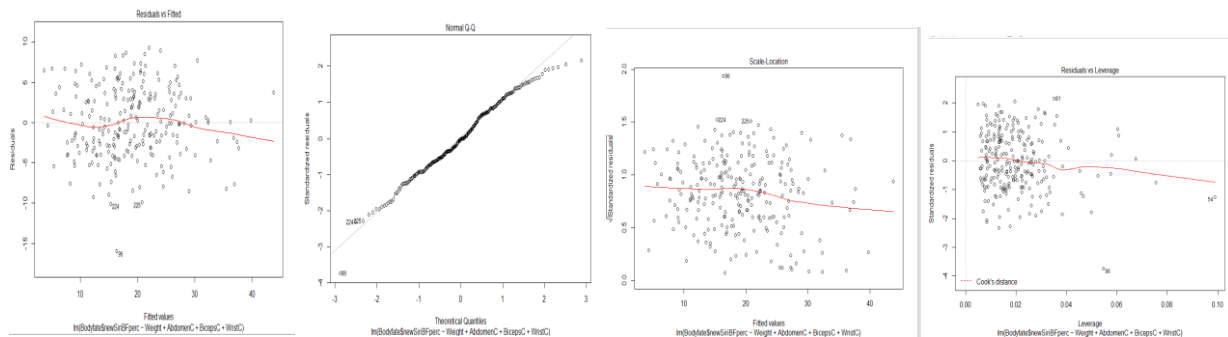
AiC model



Shapiro-wilk normality test

data: lm.aic.p\$residuals  
w = 0.99289, p-value = 0.2775

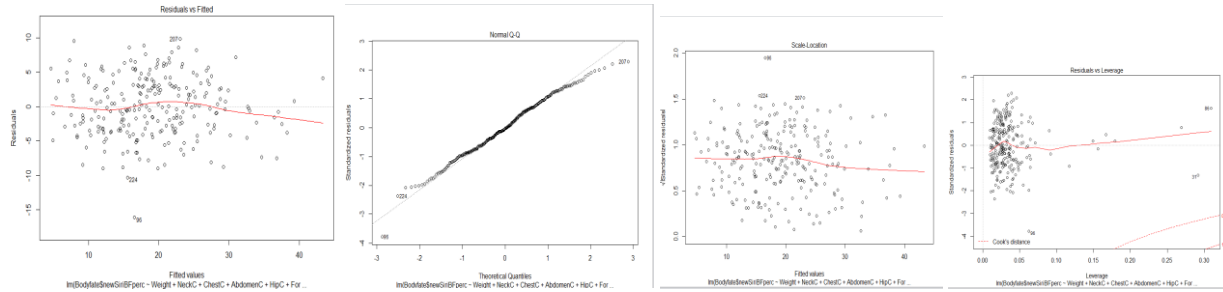
BIC model



Shapiro-wilk normality test

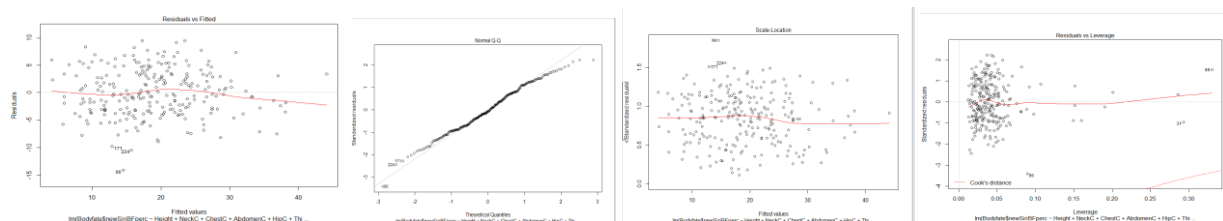
data: lm.aic.p\$residuals  
w = 0.98886, p-value = 0.05095

Cp



```
data: lm.bic.p$residuals
w = 0.9889, p-value = 0.0517
```

Adjusted R<sup>2</sup>



```
data: lm.R2.adj$residuals
W = 0.99292, p-value = 0.2809
```

A model Vif

Height	NeckC	ChestC	AbdomenC	HipC	BicepsC	WristC
1.456085	3.482255	7.188742	8.272218	5.599539	2.618959	2.419499

B model Vif:

AbdomenC	weight	wristC	BicepsC
4.298628	7.529088	2.146810	2.657965

ACI model VIF

Height	NeckC	ChestC	AbdomenC	HipC	ForearmC	WristC
1.456617	3.509739	7.112527	8.379800	5.333024	2.041229	2.457565

BIC MODEL VIF

Weight	AbdomenC	WristC
5.729221	4.263829	2.127194

R^2 adjusted

Height	NeckC	ChestC	AbdomenC	HipC	ThighC	AnkleC	BicepsC	ForearmC	WristC
1.495249	3.705737	7.463192	8.519283	9.545111	5.474892	1.654904	3.271572	2.330677	2.738371

Cp model VIF  
 weight NeckC ChestC AbdomenC HipC ForearmC AnkleC wristC  
 18.225219 3.793135 8.186659 7.965648 9.355490 2.129133 1.723900 2.6627  
 70

From this VIF, we see the model BIC's vif is less than others (the biggest one in the BIC model), and the predictor is less than others. Also from the plot, we can also see the normal distribution and others are good, I will choose the predictor weight, AbdomenC, wristC

```
> lmfinal=lm(Bodyfate$newSiriBFperc ~ weight+AbdomenC+wristC,data=Bodyfate)
> summary(lmfinal)
```

Call:  
 lm(formula = Bodyfate\$newSiriBFperc ~ weight + AbdomenC + wristC,  
 data = Bodyfate)

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	-25.23010	6.90997	-3.651	0.000319	***
weight	-0.09985	0.02483	-4.022	7.69e-05	***
AbdomenC	0.97919	0.05673	17.261	< 2e-16	***
wristC	-1.55701	0.44748	-3.479	0.000594	***

---  
 The regression model is  $\text{Bodypercent} = -25.23010 + 0.97919 \text{ AbdomenC} - 0.09985 \text{ weight} - 1.55701 \text{ wristC}$

#### The interpretation of the model

Holding other variables constant, when weight increase 1 unit, the bodypercent decrease -0.09985 and this variable's sd is 0.02483

Holding other variables constant, when AbdomenC increase 1 unit, the bodypercent increase 0.97919 and this variable's sd is 0.05673

Holding other variables constant, when wristC increase 1 unit, the bodypercent decrease -1.55701 and this variable's sd is 0.44748

The intercept of the model has no real meaning in the real life, because the bodypercent can not be negative, and weight can not be 0

For 95% CI

```
> predictors=data.frame(weight=mean(bodyfat$weight),AbdomenC=mean(bodyfat$AbdomenC),wristC=mean(bodyfat$wristC))
> PI=predict(lm.bic.p,new=predictors,interval='prediction',level=0.95)
> PI
```

	fit	lwr	upr
1	18.9852	9.985557	27.98484

(First compute the mean for each predictors, and use the prediction to predict the response of man average bodypercent)