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# LAMPIRAN

**Lampiran 1.** Data Indeks Pembangunan Manusia dan Faktor-Faktor yang Mempengaruhi di Provinsi Sulawesi Selatan Tahun 2010-2019

Kab/Kota	Tahun	IPM	RLS	HLS	PPP	AHH
Kep. Selayar	2010	62,15	6,66	11,58	7,311	67,40
Kep. Selayar	2011	62,53	6,74	11,68	7,423	67,43
Kep. Selayar	2012	62,87	6,82	11,78	7,506	67,47
Kep. Selayar	2013	63,16	6,90	11,88	7,564	67,49
Kep. Selayar	2014	63,66	7,10	11,98	7,656	67,50
Kep. Selayar	2015	64,32	7,16	12,29	7,793	67,70
Kep. Selayar	2016	64,95	7,17	12,44	8,123	67,76
Kep. Selayar	2017	65,39	7,18	12,45	8,436	67,82
Kep. Selayar	2018	66,04	7,63	12,48	8,666	67,47
Kep. Selayar	2019	67,38	7,88	12,65	9,028	68,46
Bulukumba	2010	62,73	6,58	10,79	9,022	66,15
Bulukumba	2011	63,36	6,59	11,21	9,127	66,23
Bulukumba	2012	63,82	6,61	11,56	9,136	66,31
Bulukumba	2013	64,27	6,63	11,91	9,145	66,39
Bulukumba	2014	65,24	6,66	12,31	9,618	66,43
Bulukumba	2015	65,58	6,68	12,32	9,777	66,73
Bulukumba	2016	66,46	6,86	12,64	10,040	66,84
Bulukumba	2017	67,08	7,16	12,65	10,217	66,96
Bulukumba	2018	67,70	7,43	12,91	10,331	66,31
Bulukumba	2019	68,99	7,67	13,17	10,480	67,92
⋮	⋮	⋮	⋮	⋮	⋮	⋮
⋮	⋮	⋮	⋮	⋮	⋮	⋮
⋮	⋮	⋮	⋮	⋮	⋮	⋮
Palopo	2010	73,03	9,77	13,03	11,252	69,94
Palopo	2011	74,02	9,83	13,84	11,372	70,00
Palopo	2012	74,54	9,89	14,16	11,493	70,05
Palopo	2013	75,02	9,95	14,49	11,590	70,10
Palopo	2014	75,65	9,96	15,01	11,713	70,12
Palopo	2015	76,27	10,25	15,02	12,005	70,20
Palopo	2016	76,45	10,26	15,03	12,156	70,25
Palopo	2017	76,71	10,33	15,05	12,319	70,30
Palopo	2018	77,30	10,75	15,07	12,662	70,05
Palopo	2019	78,06	10,76	15,08	12,986	70,88

## Lampiran 2. Hasil *Output* Penentuan Variabel Komponen Parametrik dan Nonparametrik

```

> #Menentukan komponen parametrik dan nonparametrik
> #1 Scatterplot
> plot(X1,Y, xlab = "RLS",ylab = "IPM",abline(lm(Y~X1),col="red"))
> plot(X2,Y, xlab = "HLS",ylab = "IPM",abline(lm(Y~X2),col="red"))
> plot(X3,Y, xlab = "PPP",ylab = "IPM",abline(lm(Y~X3),col="red"))
> plot(Z1,Y, xlab = "AHH",ylab = "IPM",abline(lm(Y~Z1),col="red"))
> plot(t,Y, xlab = "Tahun",ylab = "IPM",abline(lm(Y~t),col="red"))
> #2 Koefisien Determinasi
> regresi<-function(x,y)
+ {
+   X<-cbind(1,x)
+   Beta<-(solve(t(X)%*%X))%*%(t(X)%*%y)
+   ytopi<-X%*%Beta
+   ybar<-sum(Y)/N
+   JKE<-t(y-ytopi)%*%(y-ytopi)
+   JKT<-t(y-ybar)%*%(y-ybar)
+   JKR<-t(ytopi-ybar)%*%(ytopi-ybar)
+   Rsquare<-JKR/JKT
+   return(Rsquare)
+ }
> regresi(X1,Y)
      [,1]
[1,] 0.8718362
> regresi(X2,Y)
      [,1]
[1,] 0.7509802
> regresi(X3,Y)
      [,1]
[1,] 0.56059
> regresi(Z,Y)
      [,1]
[1,] 0.3068552
> regresi(t,Y)
      [,1]
[1,] 0.1406852
> reg<-cbind(regresi(X1,Y),regresi(X2,Y),regresi(X3,Y),regresi(Z1,Y),regresi(t,Y))
> Hasil=matrix(0,5,1)
> keputusan=matrix(0.5,5,1)
> for(i in 1:5){
+   if((reg[i])>keputusan[i]){
+     Hasil[i]="Parametrik"
+   }else
+     Hasil[i]="Non Parametrik"
+ }
> Hasil
      [,1]
[1,] "Parametrik"
[2,] "Parametrik"
[3,] "Parametrik"
[4,] "Non Parametrik"
[5,] "Non Parametrik"

```

**Lampiran 3.** Hasil *Output* Basis *B-spline* Optimal pada Orde Kuadratik dengan Dua Titik Knot  $u_1 = 2012$  dan  $u_2 = 2017$

No	$B_{-2,3}$	$B_{-1,3}$	$B_{0,3}$	$B_{1,3}$	$B_{2,3}$
1	0	0	0	0	0
2	0.25	0.67857143	0.07142857	0	0
3	0	0.71428571	0.28571429	0	0
4	0	0.45714286	0.51428571	0.02857143	0
5	0	0.25714286	0.62857143	0.11428571	0
6	0	0.11428571	0.62857143	0.25714286	0
7	0	0.02857143	0.51428571	0.45714286	0
8	0	0	0.28571429	0.71428571	0
9	0	0	0.07142857	0.67857143	0.25
10	0	0	0	0	1
11	0	0	0	0	0
12	0.25	0.67857143	0.07142857	0	0
13	0	0.71428571	0.28571429	0	0
14	0	0.45714286	0.51428571	0.02857143	0
15	0	0.25714286	0.62857143	0.11428571	0
16	0	0.11428571	0.62857143	0.25714286	0
17	0	0.02857143	0.51428571	0.45714286	0
18	0	0	0.28571429	0.71428571	0
19	0	0	0.07142857	0.67857143	0.25
20	0	0	0	0	1
⋮	⋮	⋮	⋮	⋮	⋮
⋮	⋮	⋮	⋮	⋮	⋮
⋮	⋮	⋮	⋮	⋮	⋮
231	0	0	0	0	0
232	0.25	0.67857143	0.07142857	0	0
233	0	0.71428571	0.28571429	0	0
234	0	0.45714286	0.51428571	0.02857143	0
235	0	0.25714286	0.62857143	0.11428571	0
236	0	0.11428571	0.62857143	0.25714286	0
237	0	0.02857143	0.51428571	0.45714286	0
238	0	0	0.28571429	0.71428571	0
239	0	0	0.07142857	0.67857143	0.25
240	0	0	0	0	1

**Lampiran 4.** Sintaks Program R untuk *B-Spline* Optimal pada Orde Kuadratik dengan Dua Titik Knot

```
#Aplikasi Model Semiparametrik pada Data IPM
Data<-read.csv(file.choose(), header = TRUE)

#Mendefinisikan variabel penelitian
Y<-Data[,3]
X1<-Data[,4] #parametrik
X2<-Data[,5] #parametrik
X3<-Data[,7] #parametrik
Z1<-Data[,6] #nonparametrik
t<-Data[,2]
N<-length(Y)

#Mendekati Komponen Nonparametrik dengan B-spline Koefisien Bervariasi#
B2<-splines2::bSpline(t,knots = c(2012,2017), degree = 3L)
VC2<-Z1*B2
A2<-matrix(0, ncol = 7, nrow = N)
A2[,1]<-t
A2[,2]<-t^2
A2[,3:7]<-B2

TVC2<-(t(VC2)%*%VC2)
VC2INV<-mpi(TVC2)
VC2Y<-t(VC2)%*%Y
Alpa2<-VC2INV%*%VC2Y
flamda2<-VC2%*%Alpa2
MSE2<-(t(Y-flamda2)%*%(Y-flamda2))/N
Aknot2<-(A2%*%mpi(t(A2)%*%A2)%*%t(A2))
GCV2<-MSE2/(1-((1/N)*sum(diag(Aknot2))))^2
GCV2

#Estimasi model Semiparametrik dengan Koefisien Bervariasi#
SVC<-matrix(0, ncol=9, nrow = N)
SVC[,1]<-1
SVC[,2]<-X1
SVC[,3]<-X2
SVC[,4]<-X3
SVC[,5:9]<-VC2
#Estimasi
TSVC<-t(SVC)%*%SVC
SVCINV<-solve(TSVC)
SVCINV
SVCY<-t(SVC)%*%Y
Betaalpa<-SVCINV%*%SVCY
Betaalpa
```