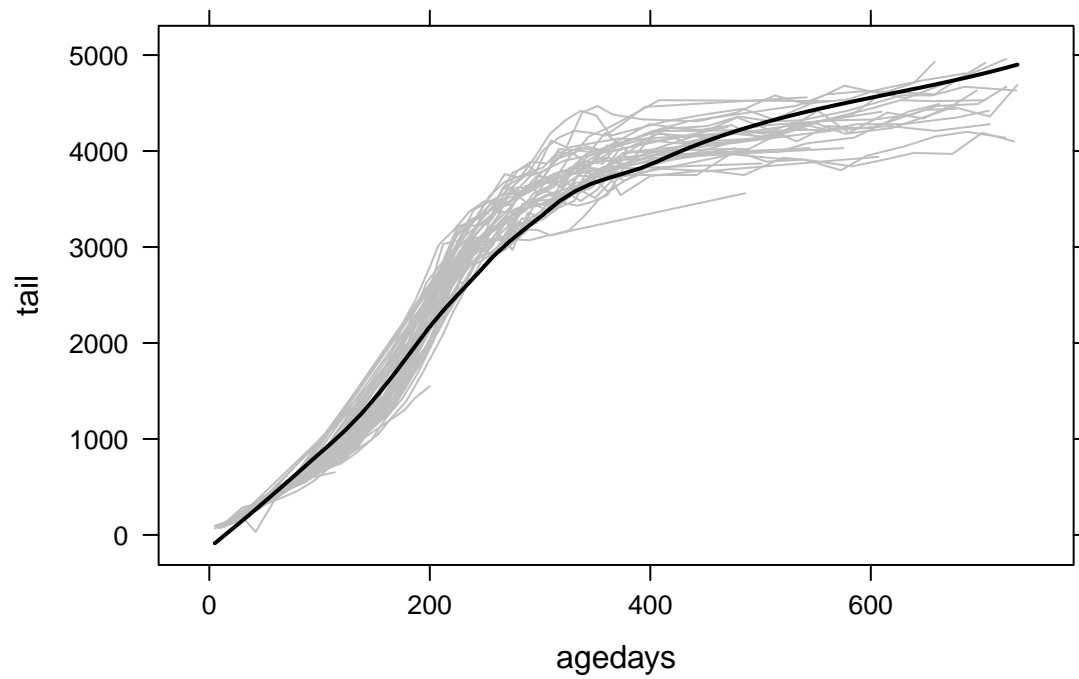


Wallaby tail data

	animal	sex	agedays	tail
1	45	1	14	93
2	45	1	28	185
3	45	1	49	307
4	45	1	69	454
5	45	1	83	568
6	45	1	97	648

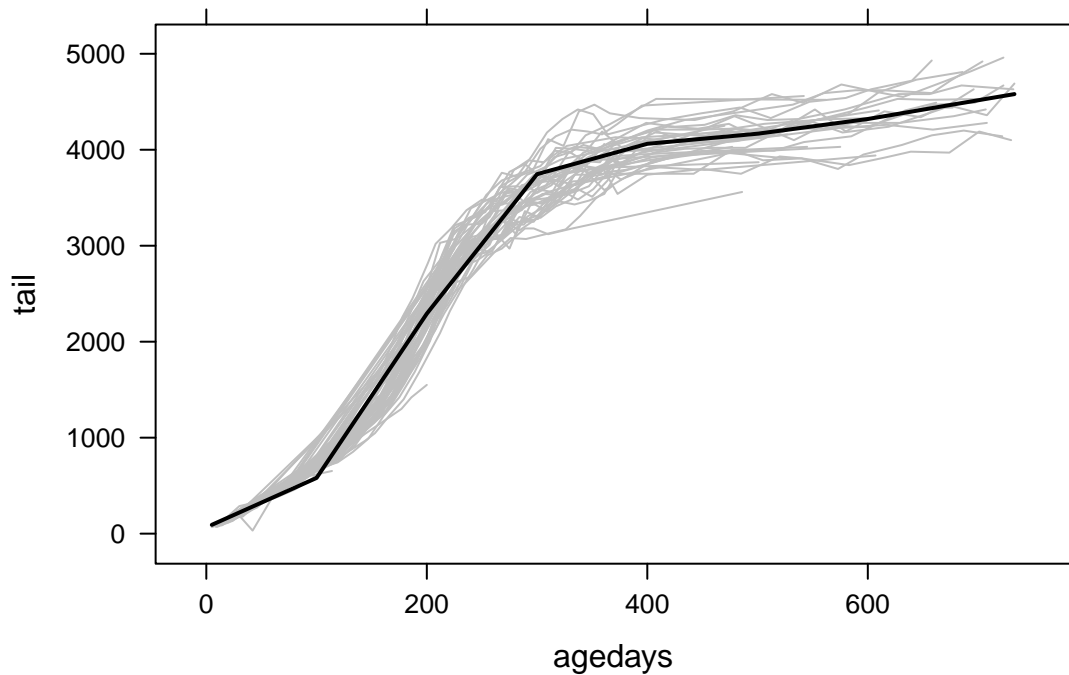


Linear Piecewise

```
> knot <- function(t, x, degree = 1) {
+   (x - t)^degree * (x > t)
+ }
> knots <- function(t, x, degree = 1) {
+   names(t) <- paste("knot", t, sep = ".")
+   sapply(t, knot, x = x, degree = degree)
+ }
> m.l1 <- lme(tail ~ agedays + knots((1:6) * 100, agedays), random = ~agedays |
+   animal, data = wal)
> summary(m.l1)
```

```
...
                Value Std.Error   DF   t-value p-value
(Intercept)      66.06923 15.578716 1257    4.24099  0.0000
agedays           5.14534  0.227726 1257   22.59448  0.0000
knots...knot.100 11.96904  0.346943 1257   34.49860  0.0000
knots...knot.200 -2.57741  0.315231 1257   -8.17628  0.0000
knots...knot.300 -11.37429 0.386183 1257  -29.45308  0.0000
knots...knot.400 -2.11200  0.535636 1257   -3.94298  0.0001
knots...knot.500  0.48214  0.658743 1257    0.73190  0.4644
knots...knot.600  0.42001  0.706652 1257    0.59437  0.5524
...

```



```

> m.l2 <- lme(tail ~ agedays + knots((1:4) * 100, agedays), random = ~agedays |
+   animal, data = wal)
> ml.l1 <- update(m.l1, method = "ML")
> ml.l2 <- update(m.l2, method = "ML")
> anova(ml.l1, ml.l2)

```

	Model	df	AIC	BIC	logLik	Test	L.Ratio	p-value
ml.l1	1	12	17237.59	17299.99	-8606.795			
ml.l2	2	10	17237.35	17289.35	-8608.673	1 vs 2	3.755532	0.1529

```

> m.l3 <- lme(tail ~ agedays + knots(c(50, 100, 150, 200, 300,
+   400), agedays), random = ~agedays | animal, data = wal)
> summary(m.l3)

```

```

...

```

	Value	Std.Error	DF	t-value	p-value
(Intercept)	18.170507	19.205429	1257	0.946113	0.3443
agedays	6.402426	0.501652	1257	12.762672	0.0000
knots...knot.50	1.055724	0.788990	1257	1.338070	0.1811
knots...knot.100	1.895643	0.709919	1257	2.670224	0.0077
knots...knot.150	15.056135	0.680427	1257	22.127467	0.0000
knots...knot.200	-11.422528	0.470835	1257	-24.260131	0.0000
knots...knot.300	-9.635222	0.306205	1257	-31.466547	0.0000
knots...knot.400	-2.005982	0.242039	1257	-8.287855	0.0000

```

...

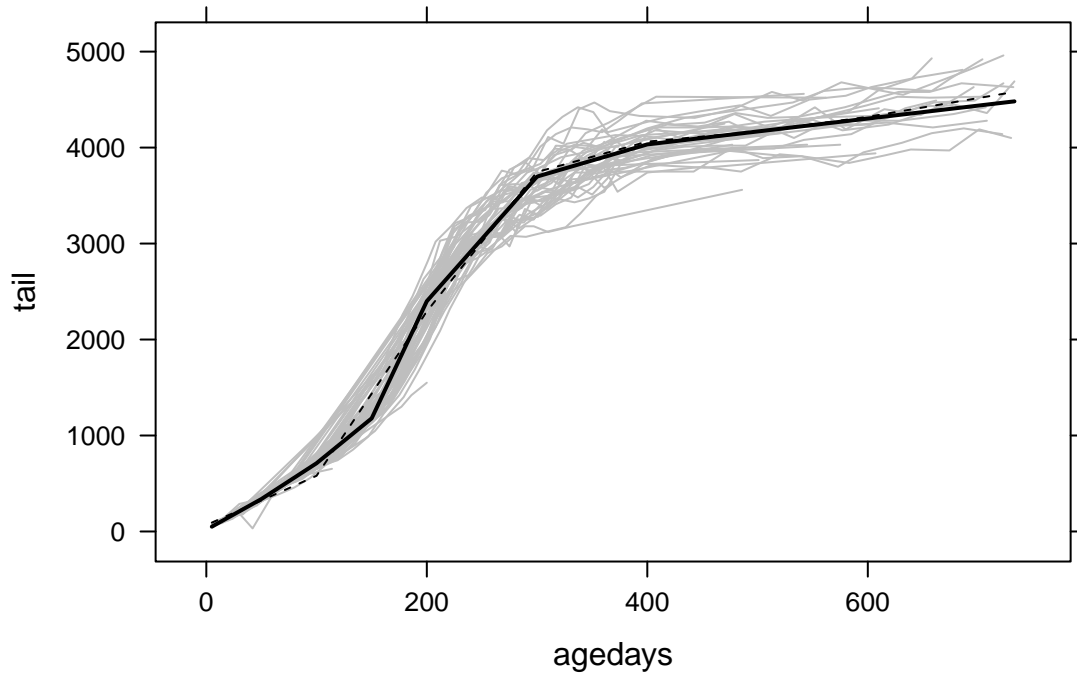
```

```

> ml.l3 <- update(m.l3, method = "ML")
> anova(ml.l2, ml.l3)

```

	Model	df	AIC	BIC	logLik	Test	L.Ratio	p-value
ml.l2	1	10	17237.35	17289.35	-8608.673			
ml.l3	2	12	16802.08	16864.48	-8389.037	1 vs 2	439.2709	<.0001



Fitted/Predicted Values

`fitted` gives the fitted (or predicted) values for the values in the data set

```
> fit <- fitted(m.l3, level = 0:1)
> fit <- data.frame(id = wal$animal, age = wal$agedays, fit)
> head(fit)
```

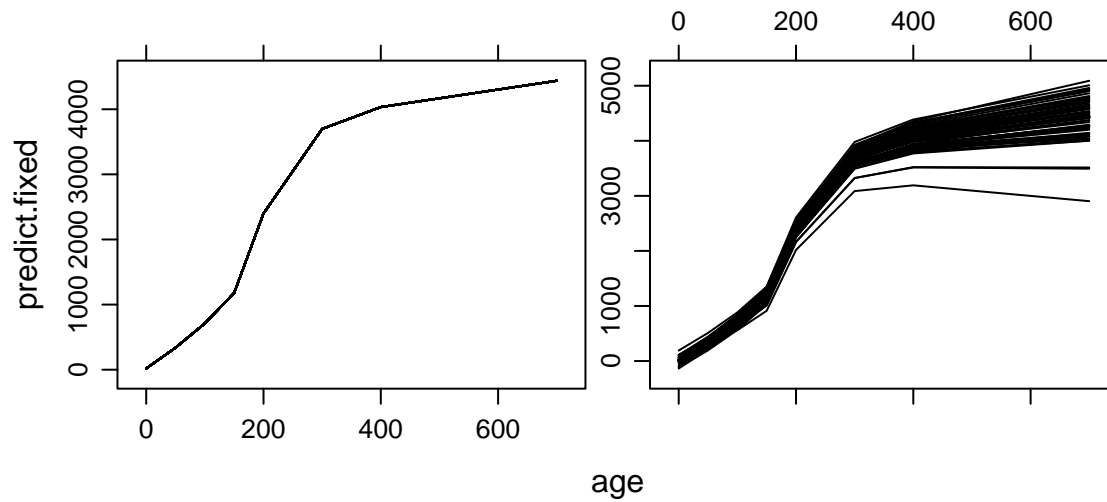
	id	age	fixed	animal
1	45	14	107.8045	43.58001
2	45	28	197.4384	135.36642
3	45	49	331.8894	273.04605
4	45	69	479.9966	424.22825
5	45	83	584.4107	530.79481
6	45	97	688.8248	637.36136

`predict` will give the fitted values for values not in the data set

```
> newd <- expand.grid(agedays = c(0, 50, 100, 150, 200, 300, 400,
+ 500, 600, 700), animal = c(44, 45))
> data.frame(agedays = newd$agedays, predict(m.l3, newdata = newd,
+ level = 0:1))
```

	agedays	animal	predict.fixed	predict.animal
1	0	44	18.17051	34.37618
2	50	44	338.29180	328.34709
3	100	44	711.19929	675.10420
4	150	44	1178.88892	1116.64345
5	200	44	2399.38530	2310.98944
6	300	44	3698.12521	3557.42859
7	400	44	4033.34295	3840.34556
8	500	44	4167.96247	3922.66433
9	600	44	4302.58200	4004.98309
10	700	44	4437.20153	4087.30185
11	0	45	18.17051	-48.20640
12	50	45	338.29180	279.60222
13	100	45	711.19929	660.19705
14	150	45	1178.88892	1135.57401
15	200	45	2399.38530	2363.75772
16	300	45	3698.12521	3677.87230
17	400	45	4033.34295	4028.46471
18	500	45	4167.96247	4178.45890
19	600	45	4302.58200	4328.45309
20	700	45	4437.20153	4478.44729

```
> newd <- expand.grid(agedays = c(0, 50, 100, 150, 200, 300, 400,
+   500, 600, 700), animal = levels(wal$animal))
> pred <- data.frame(age = newd$agedays, predict(m.l3, newd, level = 0:1))
> p1 <- xyplot(predict.fixed ~ age, group = animal, type = "l",
+   par.settings = black, data = pred)
> p2 <- xyplot(predict.animal ~ age, group = animal, type = "l",
+   par.settings = black, data = pred)
> plot(c(p1, p2))
```



Polynomial

```
> m.p2a <- lme(tail ~ agedays + I(agedays^2), random = ~agedays |
+   animal, data = wal, method = "ML")
> summary(m.p2a)
```

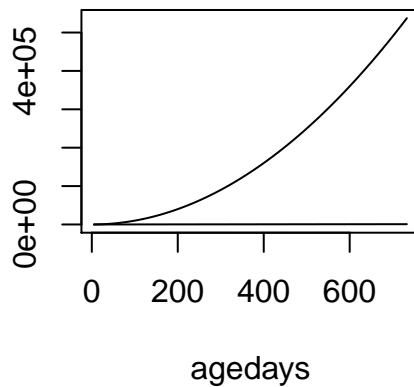
```
...
              Value Std.Error   DF   t-value p-value
(Intercept) -539.1802 21.461658 1262 -25.12295     0
agedays      17.2507  0.179214 1262  96.25781     0
I(agedays^2) -0.0147  0.000271 1262 -54.32359     0
```

```
...
```

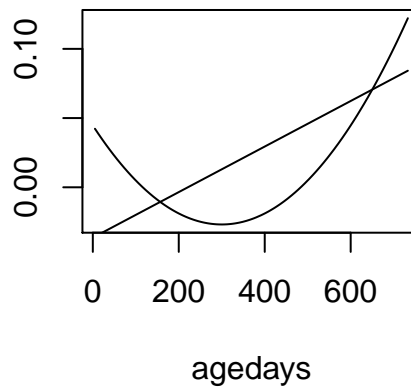
```
> poly2 <- poly(wal$agedays, 2)
> head(poly2)
```

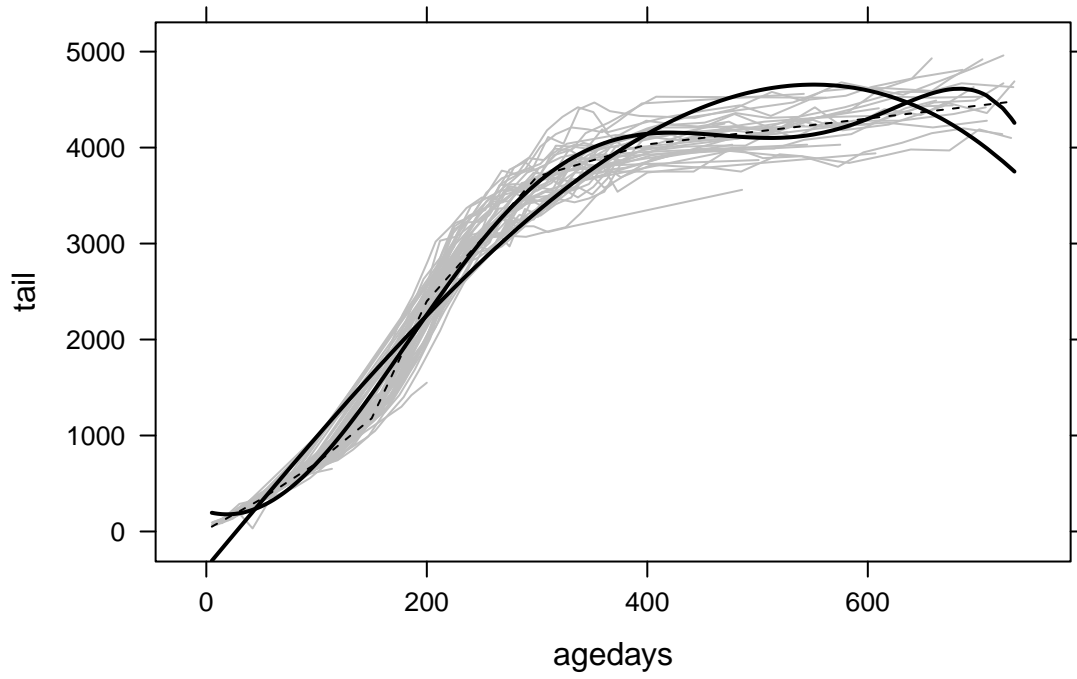
```
              1          2
[1,] -0.03407978 0.038154803
[2,] -0.03177639 0.031946559
[3,] -0.02832130 0.023218274
[4,] -0.02503074 0.015557249
[5,] -0.02272734 0.010572793
[6,] -0.02042395 0.005899847
```

By hand



Using poly





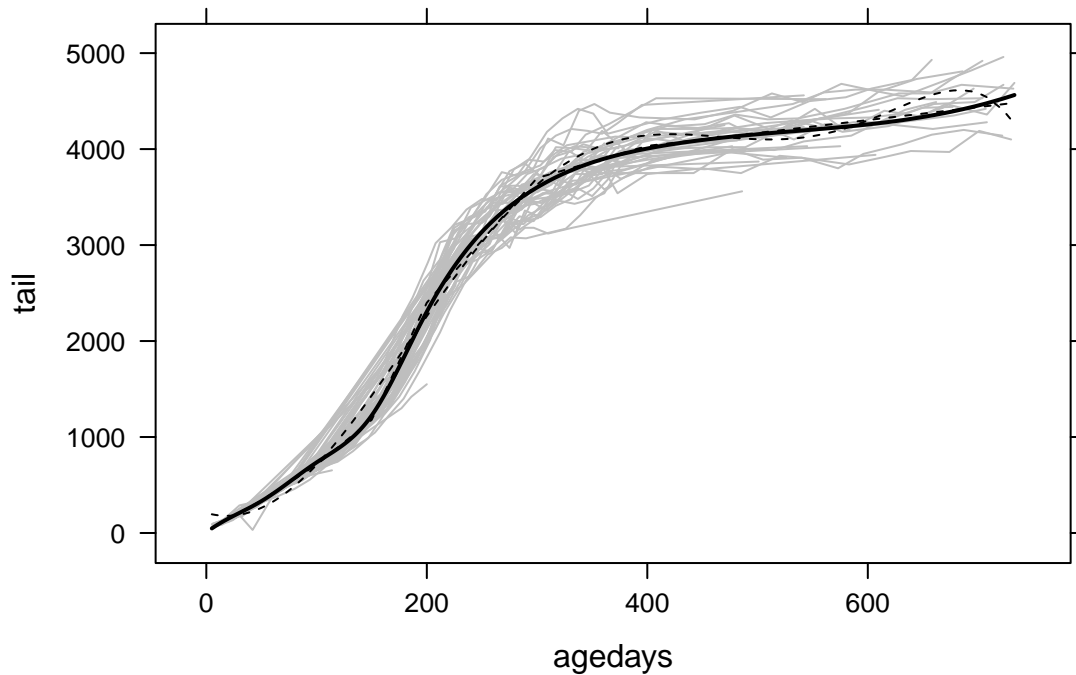
Cubic Spline

```
> m.s1 <- lme(tail ~ poly(agedays, 3) + knots(c(50, 100, 150, 200,
+ 300, 400), agedays, 3), random = ~agedays | animal, data = wal)
> summary(m.s1)
```

...

	Value	Std.Error	DF	t-value	p-value
(Intercept)	27814.1	21387.6	1255	1.300476	0.1937
poly(agedays, 3)1	1790286.7	1445445.2	1255	1.238571	0.2157
poly(agedays, 3)2	1090960.2	921948.2	1255	1.183320	0.2369
poly(agedays, 3)3	212536.5	186719.2	1255	1.138268	0.2552
knots...knot.50	0.0	0.0	1255	-1.537660	0.1244
knots...knot.100	0.0	0.0	1255	6.323144	0.0000
knots...knot.150	0.0	0.0	1255	-16.275667	0.0000
knots...knot.200	0.0	0.0	1255	21.178691	0.0000
knots...knot.300	0.0	0.0	1255	-4.210058	0.0000
knots...knot.400	0.0	0.0	1255	-3.643160	0.0003

...



B-spline

```
> library(splines)
> m.bs10 <- lme(tail ~ bs(agedays, 10), random = ~agedays | animal,
+             data = wal)
> summary(m.bs10)
```

```
...
              Value Std.Error   DF  t-value p-value
(Intercept)    54.410  30.34208 1254   1.79321  0.0732
bs(agedays, 10)1  82.833  50.49721 1254   1.64035  0.1012
bs(agedays, 10)2 212.912  30.07494 1254   7.07939  0.0000
bs(agedays, 10)3 612.171  39.92065 1254  15.33469  0.0000
bs(agedays, 10)4 887.714  34.78825 1254  25.51764  0.0000
bs(agedays, 10)5 2109.383  38.03162 1254  55.46392  0.0000
bs(agedays, 10)6 3270.694  38.19814 1254  85.62442  0.0000
bs(agedays, 10)7 3685.231  42.28260 1254  87.15715  0.0000
bs(agedays, 10)8 4316.317  61.54901 1254  70.12813  0.0000
bs(agedays, 10)9 4023.968  75.16492 1254  53.53518  0.0000
bs(agedays, 10)10 4539.785  77.16347 1254  58.83335  0.0000
...
```

