

## Ejercicio 2

Analice los datos `Machines` del ejemplo guía 3 (ec. 4) a través de la función `lme`.

### Modelo

un modelo para la  $k$ -ésima observación en el nivel  $i$  de efecto fijo  $A$  y el nivel  $j$  de efecto aleatorio  $B$  es

$$y_{ijk} = \mu + \alpha_i + b_j + (ab)_{ij} + \epsilon_{ijk}, \quad (4)$$

donde:

$$b_j \sim N(0, \sigma_b^2),$$

$$(\alpha b)_{ij} \sim N(0, \sigma_{\alpha\beta}^2),$$

$$\epsilon_{ijk} \sim N(0, \sigma^2),$$

y todas las v.a. son mutuamente independientes.

Además,  $\mu$  es la media poblacional global,  $\alpha_i$  son los  $I$  efectos fijos para el factor  $A$ , y  $b_j$  representan los  $J$  efectos aleatorios para el factor  $B$ .

Por otro lado,  $(\alpha b)_{ij}$  son las  $IJ$  interacciones.

### Datos

- El marco de datos `Machines`, del paquete `nlme`, contiene datos de un experimento industrial que compara 3 tipos de máquinas diferentes.
- El objetivo del experimento es determinar qué tipo de máquina daba como resultado la mayor productividad de los trabajadores.
- Se seleccionaron al azar 6 trabajadores para participar en la prueba, y cada trabajador operó cada máquina 3 veces (presumiblemente después de un período apropiado de capacitación diseñado para eliminar cualquier "efecto de aprendizaje").

**(a) Intente encontrar el modelo más apropiado, teniendo cuidado de examinar los gráficos de verificación del modelo apropiado.**

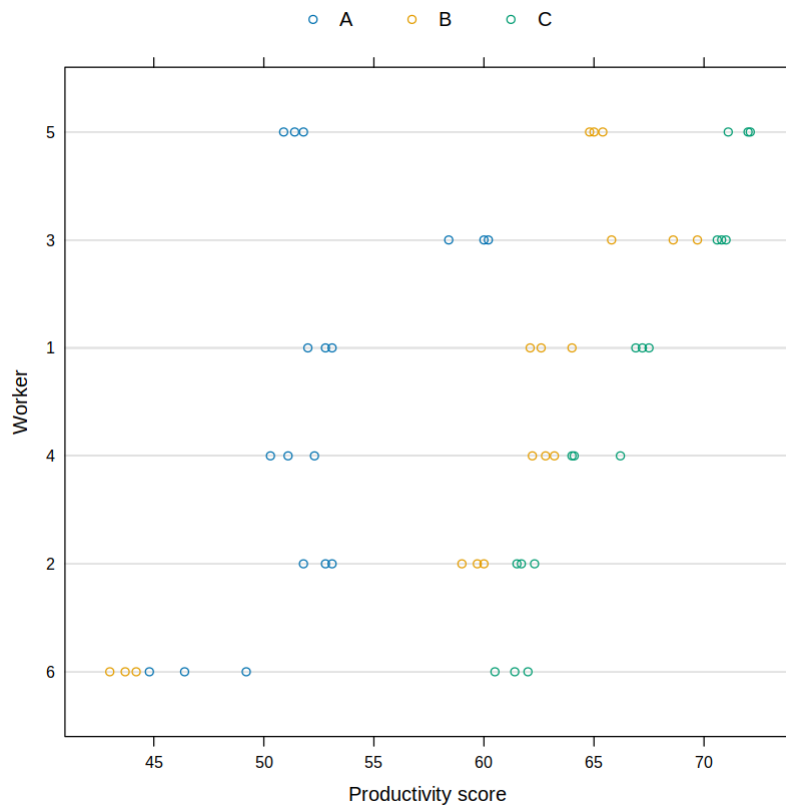
```
In [1]: library(nlme)
attach(Machines)
Machines
```

A nffGroupedData: 54 × 3

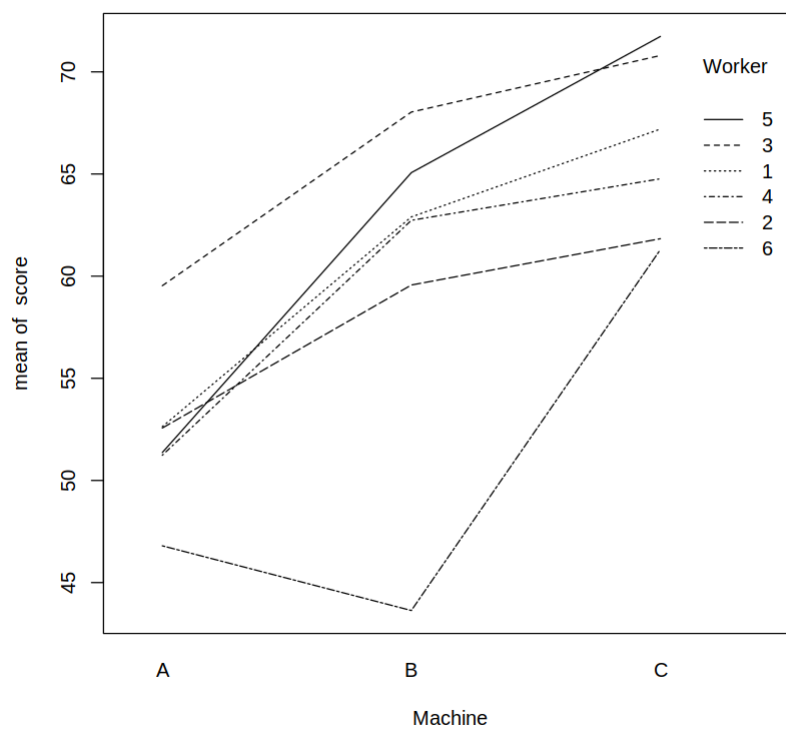
	Worker	Machine	score
	<ord>	<fct>	<dbl>
1	1	A	52.0
2	1	A	52.8
3	1	A	53.1
4	2	A	51.8
5	2	A	52.8
6	2	A	53.1
7	3	A	60.0
8	3	A	60.2
9	3	A	58.4
10	4	A	51.1
11	4	A	52.3
12	4	A	50.3
13	5	A	50.9
14	5	A	51.8
15	5	A	51.4
16	6	A	46.4
17	6	A	44.8
18	6	A	49.2
19	1	B	62.1
20	1	B	62.6
21	1	B	64.0
22	2	B	59.7
23	2	B	60.0
24	2	B	59.0
25	3	B	68.6
26	3	B	65.8
27	3	B	69.7
28	4	B	63.2
29	4	B	62.8
30	4	B	62.2
31	5	B	64.8
32	5	B	65.0

	Worker	Machine	score
	<ord>	<fct>	<dbl>
33	5	B	65.4
34	6	B	43.7
35	6	B	44.2
36	6	B	43.0
37	1	C	67.5
38	1	C	67.2
39	1	C	66.9
40	2	C	61.5
41	2	C	61.7
42	2	C	62.3
43	3	C	70.8
44	3	C	70.6
45	3	C	71.0
46	4	C	64.1
47	4	C	66.2
48	4	C	64.0
49	5	C	72.1
50	5	C	72.0
51	5	C	71.1
52	6	C	62.0
53	6	C	61.4
54	6	C	60.5

In [2]: `plot(Machines)`



```
In [3]: interaction.plot( Machine, Worker, score)
```



Construimos un primer modelo sin interaccion

```
In [4]: m0 <- lme( score ~ Machine, data = Machines, random = ~ 1 | Worker )  
m0
```

Linear mixed-effects model fit by REML

Data: Machines

Log-restricted-likelihood: -143.4391

Fixed: score ~ Machine

(Intercept)	MachineB	MachineC
52.355556	7.966667	13.916667

Random effects:

Formula: ~1 | Worker

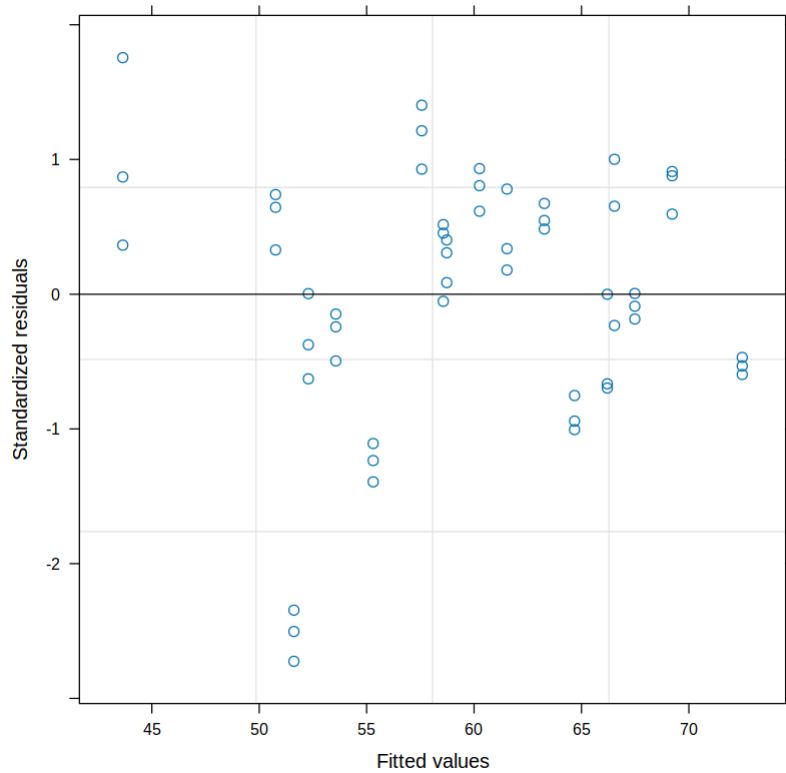
(Intercept) Residual

StdDev: 5.146552 3.161647

Number of Observations: 54

Number of Groups: 6

```
In [5]: plot(m0)
```



construimos un modelo con interaccion

```
In [6]: m1 <- lme( score ~ Machine, data = Machines, random = ~ 1 | Worker/Machine )  
m1
```

Linear mixed-effects model fit by REML

Data: Machines

Log-restricted-likelihood: -107.8438

Fixed: score ~ Machine

(Intercept)	MachineB	MachineC
52.355556	7.966667	13.916667

Random effects:

Formula: ~1 | Worker

(Intercept)

StdDev: 4.78105

Formula: ~1 | Machine %in% Worker

(Intercept) Residual

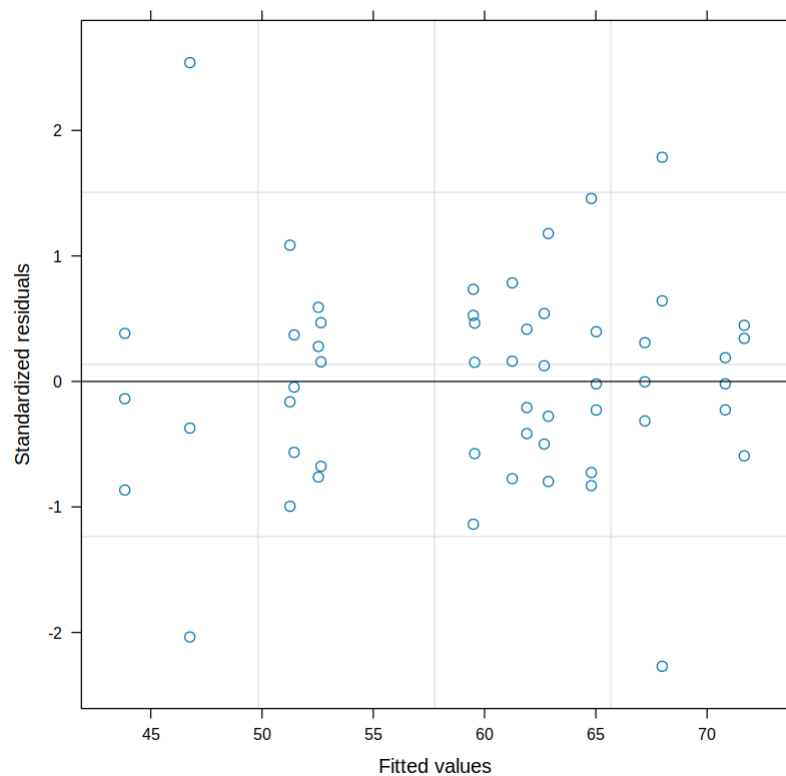
StdDev: 3.729532 0.9615771

Number of Observations: 54

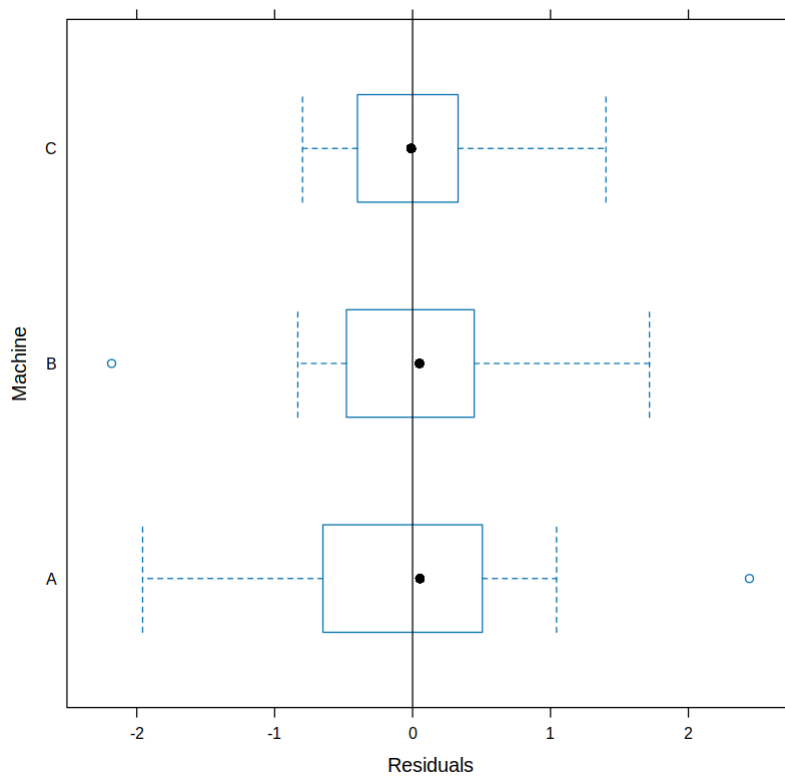
Number of Groups:

Worker	Machine %in% Worker
6	18

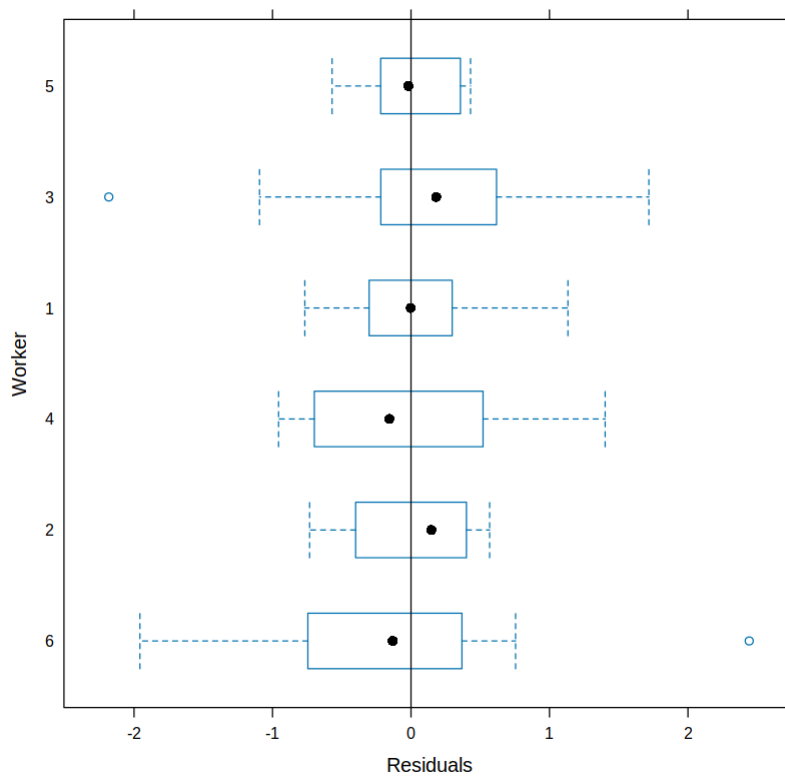
In [7]: `plot(m1)`



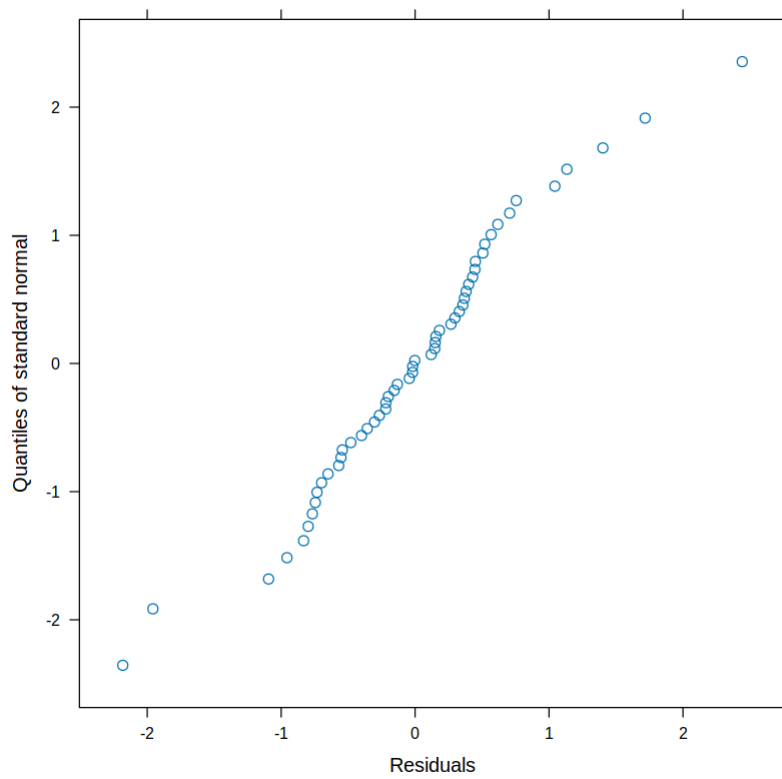
In [8]: `plot(m1, Machine~resid(.), abline=0)`



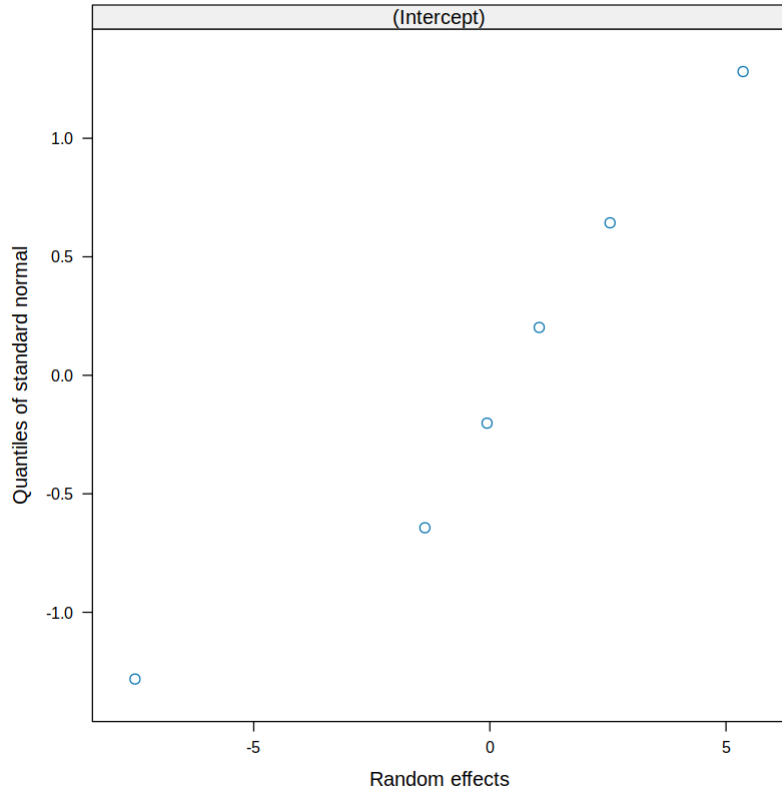
```
In [9]: plot(m1, Worker ~ resid(.), abline=0)
```



```
In [10]: qqnorm(m1, ~ resid(.))
```

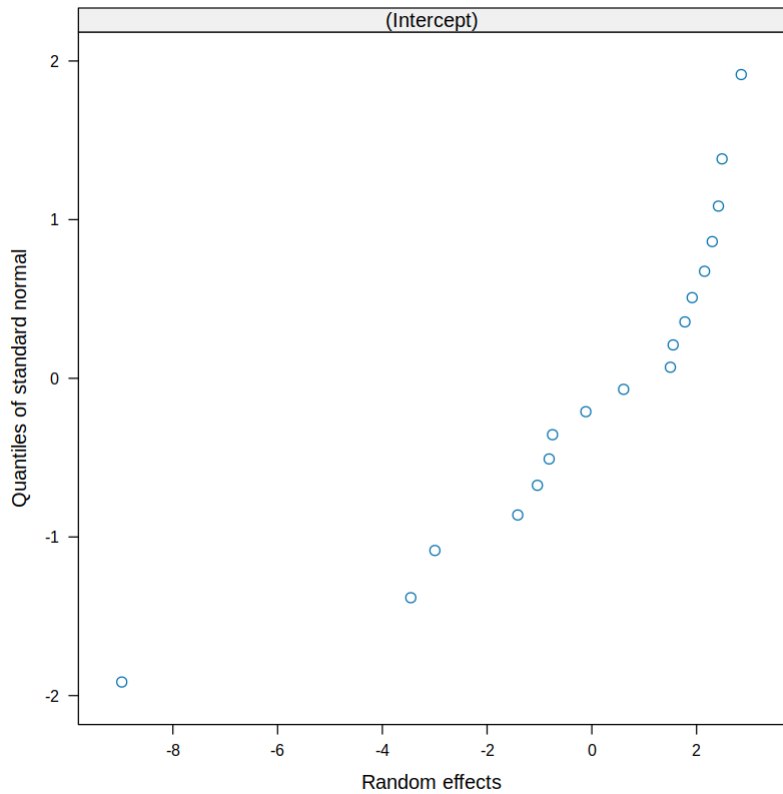


```
In [11]: qqnorm(m1,~ranef(.,level=1))
```



```
In [12]: qqnorm(m1,~ranef(.,level=2))
```





Se puede observar un dato atípico

(b) Asegúrese de probar si la interacción de la ec. 4 es apropiada.

```
In [13]: anova(m0, m1)
```

A anova.lme: 2 × 9

	call	Model	df	AIC	BIC	logLik	Test	L.Ratio	p-value
	<chr>	<int>	<dbl>	<dbl>	<dbl>	<dbl>	<fct>	<dbl>	<dbl>
<b>m0</b>	lme.formula(fixed = score ~ Machine, data = Machines, random = ~1   Worker)	1	5	296.8782	306.5373	-143.4391		NA	
<b>m1</b>	lme.formula(fixed = score ~ Machine, data = Machines, random = ~1   Worker/Machine)	2	6	227.6876	239.2785	-107.8438	1 vs 2	71.19063	3.243e-13

```
In [14]: anova(m1)
```

A anova.lme: 2 × 4

	numDF	denDF	F-value	p-value
	<int>	<dbl>	<dbl>	<dbl>
(Intercept)	1	36	773.57093	0.00000000000
Machine	2	10	20.57615	0.0002855446

In [15]: `intervals(m1)`

Approximate 95% confidence intervals

Fixed effects:

	lower	est.	upper
(Intercept)	47.314062	52.355556	57.39705
MachineB	3.116071	7.966667	12.81726
MachineC	9.066071	13.916667	18.76726

Random Effects:

Level: Worker

	lower	est.	upper
sd((Intercept))	2.249875	4.78105	10.15987

Level: Machine

	lower	est.	upper
sd((Intercept))	2.382833	3.729532	5.83734

Within-group standard error:

	lower	est.	upper
	0.7635351	0.9615771	1.2109864

Todos los criterios dan preferencia al modelo con interaccion. Particularmente, el criterio con interaccion incrementa la verosimilitud.

**(c) De manera similar, pruebe si sería apropiada una estructura de efectos aleatorios más compleja: específicamente una en la que la interacción máquina-trabajador esté correlacionada con la del trabajador.**

In [16]: `m2 <- lme( score ~ Machine, data = Machines, random = ~ Machine | Worker)  
summary(m2)`

Linear mixed-effects model fit by REML

Data: Machines

	AIC	BIC	logLik
	228.3112	247.6295	-104.1556

Random effects:

Formula: ~Machine | Worker

Structure: General positive-definite, Log-Cholesky parametrization

	StdDev	Corr
(Intercept)	4.0792806	(Intr) MachnB
MachineB	5.8776433	0.484
MachineC	3.6898543	-0.365 0.297
Residual	0.9615766	

Fixed effects: score ~ Machine

	Value	Std.Error	DF	t-value	p-value
(Intercept)	52.35556	1.680711	46	31.150834	0.0000
MachineB	7.96667	2.420851	46	3.290854	0.0019
MachineC	13.91667	1.540100	46	9.036211	0.0000

Correlation:

	(Intr) MachnB
MachineB	0.463
MachineC	-0.374 0.301

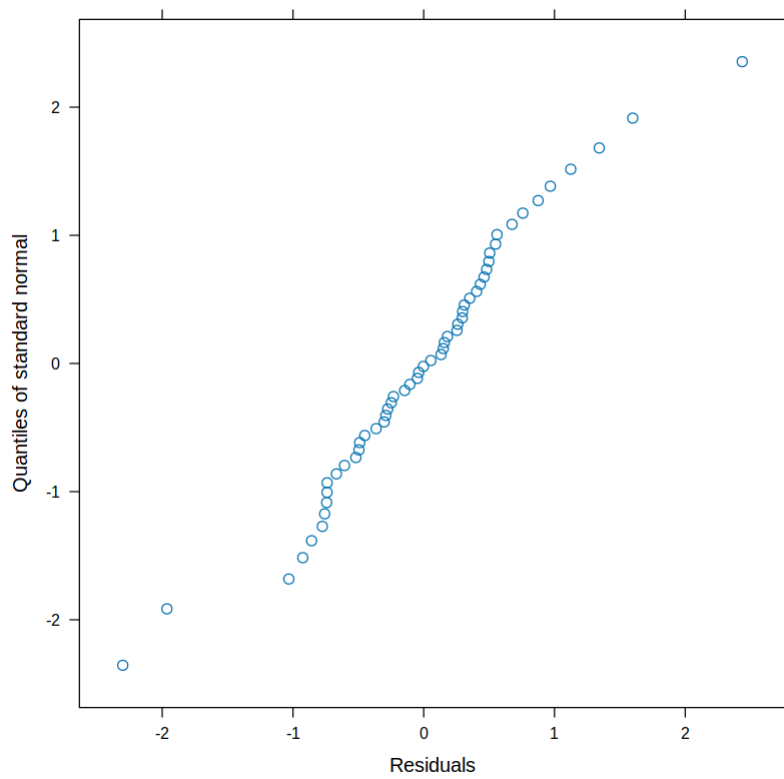
Standardized Within-Group Residuals:

	Min	Q1	Med	Q3	Max
	-2.39354008	-0.51377574	0.02690829	0.47245471	2.53338699

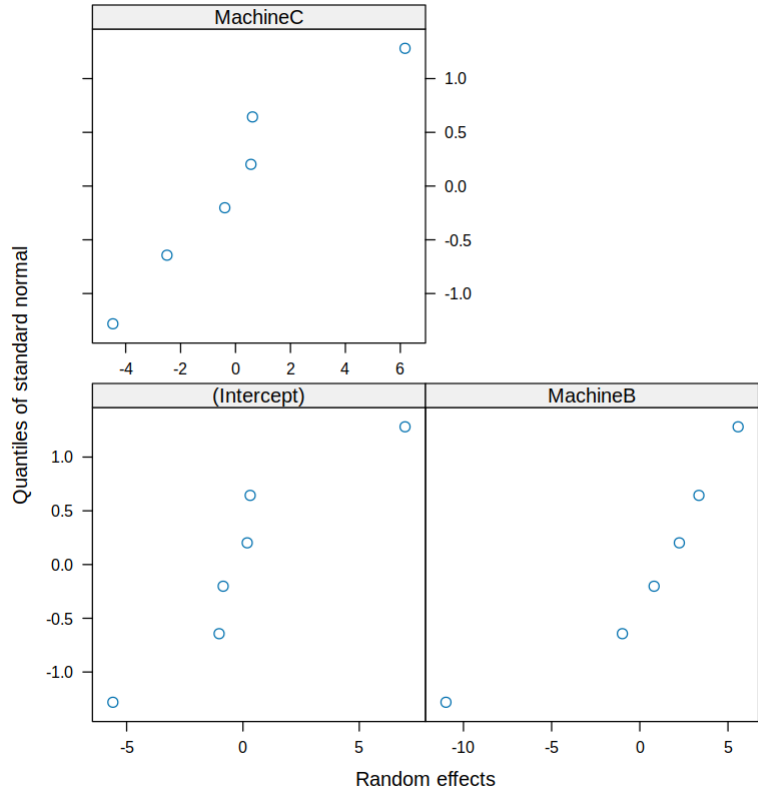
Number of Observations: 54

Number of Groups: 6

In [17]: qqnorm(m2,~resid(.))



```
In [18]: qqnorm(m2,~ranef(.,level=1))
```



```
In [19]: anova(m0,m1,m2)
```

A anova.lme: 3 × 9									
	call	Model	df	AIC	BIC	logLik	Test	L.Ratio	p-value
	<chr>	<int>	<dbl>	<dbl>	<dbl>	<dbl>	<fct>	<dbl>	<dbl>
<b>m0</b>	lme.formula(fixed = score ~ Machine, data = Machines, random = ~1   Worker)	1	5	296.8782	306.5373	-143.4391		NA	
<b>m1</b>	lme.formula(fixed = score ~ Machine, data = Machines, random = ~1   Worker/Machine)	2	6	227.6876	239.2785	-107.8438	1 vs 2	71.19063	3.243e-13
<b>m2</b>	lme.formula(fixed = score ~ Machine, data = Machines, random = ~Machine   Worker)	3	10	228.3112	247.6295	-104.1556	2 vs 3	7.37635	1.172e-01

In [20]: `anova(m2)`

A anova.lme: 2 × 4				
	numDF	denDF	F-value	p-value
	<int>	<dbl>	<dbl>	<dbl>
<b>(Intercept)</b>	1	46	2351.80626	0.000000e+00
<b>Machine</b>	2	46	41.00377	5.984291e-11

(d) Si algún dato parece particularmente problemático en los gráficos de verificación, repita el análisis y vea si las conclusiones cambian.

Se eliminan los datos relacionados con el trabajador 6

In [27]: `Machines <- Machines[-(34:36),]`

In [28]: `m3 <- lme( score ~ Machine, data = Machines, random = ~ 1 | Worker/Machine )`  
`summary(m3)`

Linear mixed-effects model fit by REML

Data: Machines

AIC	BIC	logLik
193.0094	203.8493	-90.50468

Random effects:

Formula: ~1 | Worker  
(Intercept)

StdDev: 3.530511

Formula: ~1 | Machine %in% Worker  
(Intercept) Residual

StdDev: 2.084565 1.00592

Fixed effects: score ~ Machine

	Value	Std.Error	DF	t-value	p-value
(Intercept)	52.35556	1.690523	32	30.970030	0e+00
MachineB	10.43100	1.335769	8	7.808985	1e-04
MachineC	13.76405	1.335769	8	10.304214	0e+00

Correlation:

(Intr) MachnB

MachineB -0.346

MachineC -0.346 0.427

Standardized Within-Group Residuals:

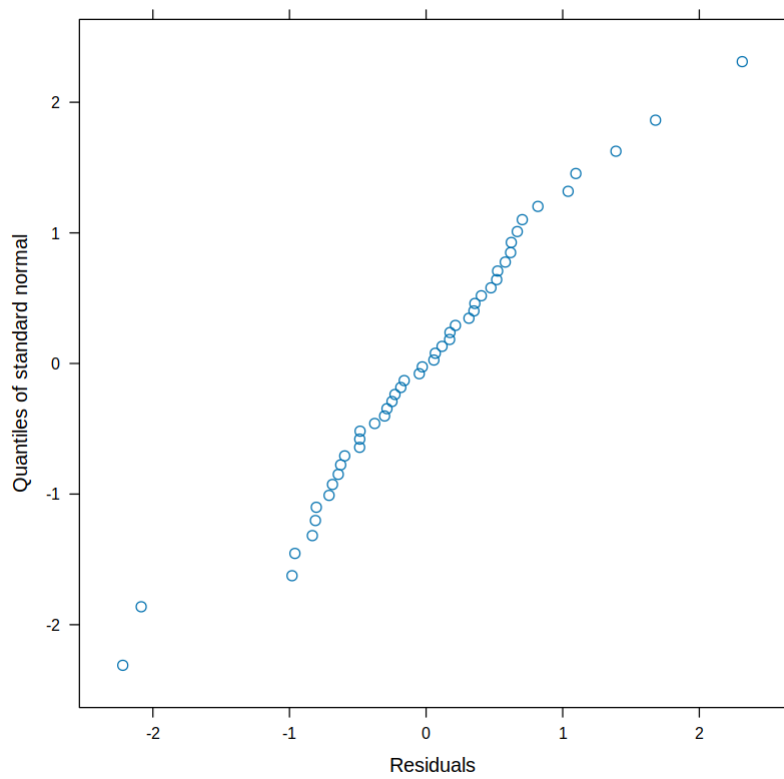
Min	Q1	Med	Q3	Max
-2.20732393	-0.51073470	0.01452005	0.51545659	2.30078469

Number of Observations: 48

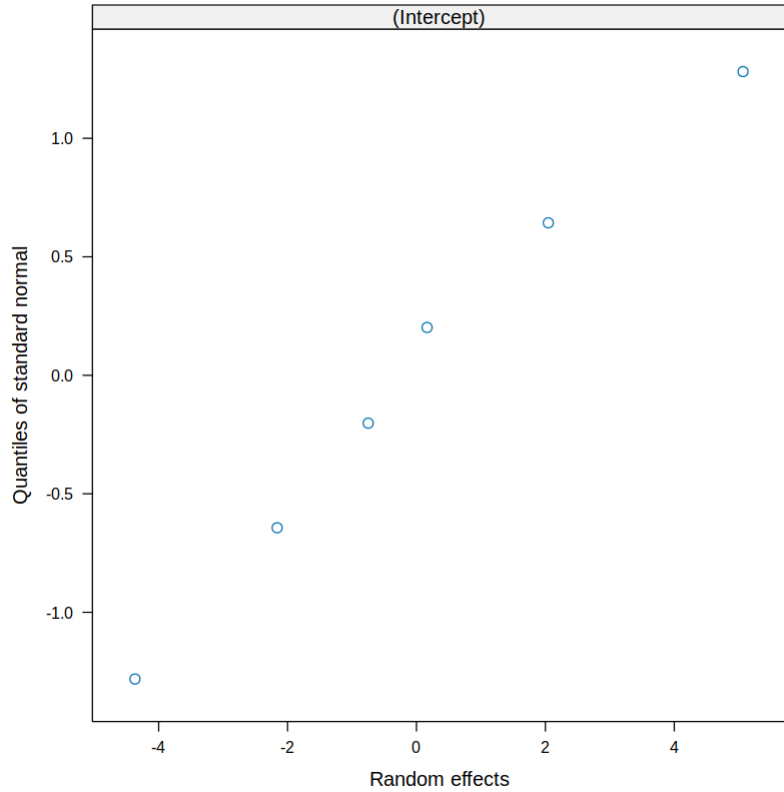
Number of Groups:

Worker	Machine %in% Worker
6	16

In [32]: qqnorm(m3,~resid(.))



```
In [29]: qqnorm(m3,~ranef(.,level=1))
```



```
In [30]: qqnorm(m3,~ranef(.,level=2))
```

