

**Obtain parameters (e.g., variances, covariances, correlations ...) at all time points given RR coefficients from a random regression analysis:**

In general, RR coefficients can be transformed to over-time estimates (time function) using the polynomials used to get those coefficients. Normally, these RR coefficients are obtained based on a range of time points. For example, ages in days: 1,3,4,6,7,8. But, we could also predict the values where we do not have time points. For example, we could predict the estimates at ages 2 and 5. However, it is important that those missing time points fall within the range of the observed range (i.e., between 1 and 8). In other words, **we should not use these RR coefficients for predicting time periods outside the observed range**. The general approach as follows (matrix multiplication):

(Legendre polynomials x regression coefficient x transpose of Legendre polynomials)

**1. Single trait:**

For a single trait, it is quite simple because we do not have covariances between multiple traits. Note: the time points that have been used to obtain those variance components should be known as well. Assume we would like to get variances and covariances (for the single trait) over time (File: 3\_RR\_coefficients\_for\_1\_traits.txt):

RR coefficients	1	2	3
1	3.5	0.8	0.5
2	0.8	1.3	0.4
3	0.5	0.4	1

The estimates were obtained based on 5 time points, then the Legendre polynomials are:

```
0.7071068 -1.2247449 1.5811388
0.7071068 -0.6123724 -0.1976424
0.7071068 0.0000000 -0.7905694
0.7071068 0.6123724 -0.1976424
0.7071068 1.2247449 1.5811388
```

Using the R code (3\_RR\_coefficients\_for\_1\_traits.R) or the function (RRcoeff\_to\_varComp\_function.R or RRcoeff\_to\_varComp\_function.py), which is flexible for any number of traits, we get:

varCovr

```
      [,1]      [,2]      [,3]      [,4]      [,5]
[1,] 4.383200 1.5719356 0.4739866 1.089353 3.4180339
[2,] 1.5719356 1.5408126 1.4041035 1.161808 0.8139272
[3,] 0.4739866 1.4041035 1.8159831 1.709625 1.0850307
[4,] 1.0893526 1.1618084 1.7096255 2.732804 4.2313441
[5,] 3.4180339 0.8139272 1.0850307 4.231344 10.2528681
```

#The additive genetic variance:

```
[1] 4.383200 1.540813 1.815983 2.732804 10.252868
```

correlation

```
      [,1]      [,2]      [,3]      [,4]      [,5]
[1,] 1.0000000 0.6048730 0.1680020 0.3147521 0.5098683
[2,] 0.6048730 1.0000000 0.8393987 0.5661811 0.2047802
[3,] 0.1680020 0.8393987 1.0000000 0.7674337 0.2514568
[4,] 0.3147521 0.5661811 0.7674337 1.0000000 0.7993766
[5,] 0.5098683 0.2047802 0.2514568 0.7993766 1.0000000
```



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From the R code (3\_RR\_coefficients\_for\_2\_traits.R):

```
> #=====#
> Phi = as.matrix(read.table("Legendre_polynomials.txt"))
> H = as.matrix(read.table("3_RR_coefficients_for_2_traits.txt", header =F))
> #=====#
> > H1 = H[1:3,1:3] # Block 1
> H1
      V1  V2  V3
[1,] 3.5 0.8 0.5
[2,] 0.8 1.3 0.4
[3,] 0.5 0.4 1.0
> H2 = H[1:3,4:6] # Block 2, same as transpose of H[4:6,1:3]
> H2
      V4  V5  V6
[1,] 0.8 0.9 1.50
[2,] 0.2 1.6 0.25
[3,] 0.5 5.0 -1.50
> H3 = H[4:6,4:6] # Block 3
> H3
      V4  V5  V6
[1,] 2.0 0.5 -2
[2,] 0.5 3.0 2
[3,] -2.0 2.0 5
>
> HH1 = Phi %*%H1 %*% t(Phi)
> HH1
      [,1]      [,2]      [,3]      [,4]      [,5]
[1,] 4.3832000 1.5719356 0.4739866 1.089353 3.4180339
[2,] 1.5719356 1.5408126 1.4041035 1.161808 0.8139272
[3,] 0.4739866 1.4041035 1.8159831 1.709625 1.0850307
[4,] 1.0893526 1.1618084 1.7096255 2.732804 4.2313441
[5,] 3.4180339 0.8139272 1.0850307 4.231344 10.2528681
> HH2 = Phi %*%H2 %*% t(Phi)
> HH2
      [,1]      [,2]      [,3]      [,4]      [,5]
[1,] -9.833141 -2.9254943 2.0643478 5.1363863 6.2906213
[2,] 3.778145 0.8209951 -0.7083495 -0.8098894 0.5163753
[3,] 7.734349 1.7073880 -1.6555339 -2.3544179 -0.3892639
[4,] 2.035473 -0.2663151 -0.7772059 0.5028001 3.5737032
[5,] -13.318485 -5.1001148 1.9266351 7.7617656 12.4052776
```

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```
> HH3 = Phi %*%H3 %*% t(Phi)
```

```
> HH3
```

```
      [,1]      [,2]      [,3]      [,4]      [,5]
[1,]  4.915872 -2.3709473 -4.864555 -2.5649514  4.527863
[2,] -2.370947  2.9304400  3.930532  0.6293299 -6.973168
[3,] -4.864555  3.9305322  6.361068  2.4270534 -7.871513
[4,] -2.564951  0.6293299  2.427053  2.8282193  1.832828
[5,]  4.527863 -6.9731678 -7.871513  1.8328278 22.139856
```

```
>
```

```
> temp1 = cbind(HH1,HH2)
```

```
> temp1
```

```
      [,1]      [,2]      [,3]      [,4]      [,5]      [,6]      [,7]      [,8]      [,9]      [,10]
[1,]  4.3832000  1.5719356  0.4739866  1.089353  3.4180339 -9.833141 -2.9254943  2.0643478  5.1363863  6.2906213
[2,]  1.5719356  1.5408126  1.4041035  1.161808  0.8139272  3.778145  0.8209951 -0.7083495 -0.8098894  0.5163753
[3,]  0.4739866  1.4041035  1.8159831  1.709625  1.0850307  7.734349  1.7073880 -1.6555339 -2.3544179 -0.3892639
[4,]  1.0893526  1.1618084  1.7096255  2.732804  4.2313441  2.035473 -0.2663151 -0.7772059  0.5028001  3.5737032
[5,]  3.4180339  0.8139272  1.0850307  4.231344 10.2528681 -13.318485 -5.1001148  1.9266351  7.7617656 12.4052776
```

```
> temp2 = cbind(t(HH2),HH3) # Make sure to transpose Block 2
```

```
> temp2
```

```
      [,1]      [,2]      [,3]      [,4]      [,5]      [,6]      [,7]      [,8]      [,9]      [,10]
[1,] -9.833141  3.7781447  7.7343487  2.0354727 -13.318485  4.915872 -2.3709473 -4.864555 -2.5649514  4.527863
[2,] -2.925494  0.8209951  1.7073880 -0.2663151 -5.100115 -2.370947  2.9304400  3.930532  0.6293299 -6.973168
[3,]  2.064348 -0.7083495 -1.6555339 -0.7772059  1.926635 -4.864555  3.9305322  6.361068  2.4270534 -7.871513
[4,]  5.136386 -0.8098894 -2.3544179  0.5028001  7.761766 -2.564951  0.6293299  2.427053  2.8282193  1.832828
[5,]  6.290621  0.5163753 -0.3892639  3.5737032 12.405278  4.527863 -6.9731678 -7.871513  1.8328278 22.139856
```

```
> varCovr = rbind(temp1,temp2)
```

```
> varCovr
```

```
      [,1]      [,2]      [,3]      [,4]      [,5]      [,6]      [,7]      [,8]      [,9]      [,10]
[1,]  4.3832000  1.5719356  0.4739866  1.0893526  3.4180339 -9.833141 -2.9254943  2.0643478  5.1363863  6.2906213
[2,]  1.5719356  1.5408126  1.4041035  1.1618084  0.8139272  3.778145  0.8209951 -0.7083495 -0.8098894  0.5163753
[3,]  0.4739866  1.4041035  1.8159831  1.7096255  1.0850307  7.734349  1.7073880 -1.6555339 -2.3544179 -0.3892639
[4,]  1.0893526  1.1618084  1.7096255  2.7328040  4.2313441  2.035473 -0.2663151 -0.7772059  0.5028001  3.5737032
[5,]  3.4180339  0.8139272  1.0850307  4.2313441 10.2528681 -13.318485 -5.1001148  1.9266351  7.7617656 12.4052776
[6,] -9.8331410  3.7781447  7.7343487  2.0354727 -13.3184853  4.915872 -2.3709473 -4.8645547 -2.5649514  4.5278634
[7,] -2.9254943  0.8209951  1.7073880 -0.2663151 -5.1001148 -2.370947  2.9304400  3.9305322  0.6293299 -6.9731678
[8,]  2.0643478 -0.7083495 -1.6555339 -0.7772059  1.9266351 -4.864555  3.9305322  6.3610679  2.4270534 -7.8715127
[9,]  5.1363863 -0.8098894 -2.3544179  0.5028001  7.7617656 -2.564951  0.6293299  2.4270534  2.8282193  1.8328278
[10,]  6.2906213  0.5163753 -0.3892639  3.5737032 12.4052776  4.527863 -6.9731678 -7.8715127  1.8328278 22.1398560
```

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```
> correlation = cov2cor(varCovr)
```

```
> correlation
```

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]	[,8]	[,9]	[,10]
[1,]	1.0000000	0.60487302	0.16800203	0.31475207	0.5098683	-2.1183434	-0.81627631	0.3909509	1.4588327	0.63857293
[2,]	0.6048730	1.00000000	0.83939874	0.56618109	0.2047802	1.3727878	0.38636630	-0.2262599	-0.3879666	0.08841037
[3,]	0.1680020	0.83939874	1.00000000	0.76743366	0.2514568	2.5886157	0.74013326	-0.4870989	-1.0388937	-0.06139045
[4,]	0.3147521	0.56618109	0.76743366	1.00000000	0.7993766	0.5553423	-0.09410769	-0.1864090	0.1808567	0.45943793
[5,]	0.5098683	0.20478021	0.25145682	0.79937657	1.0000000	-1.8759959	-0.93044531	0.2385678	1.4413893	0.82337259
[6,]	-2.1183434	1.37278780	2.58861573	0.55534233	-1.8759959	1.0000000	-0.62467658	-0.8699170	-0.6878953	0.43401591
[7,]	-0.8162763	0.38636630	0.74013326	-0.09410769	-0.9304453	-0.6246766	1.00000000	0.9103742	0.2186026	-0.86571772
[8,]	0.3909509	-0.22625994	-0.48709893	-0.18640897	0.2385678	-0.8699170	0.91037417	1.0000000	0.5722131	-0.66329357
[9,]	1.4588327	-0.38796655	-1.03889371	0.18085669	1.4413893	-0.6878953	0.21860265	0.5722131	1.0000000	0.23162088
[10,]	0.6385729	0.08841037	-0.06139045	0.45943793	0.8233726	0.4340159	-0.86571772	-0.6632936	0.2316209	1.00000000