

Gasto cardíaco por ECOTT vs Fick

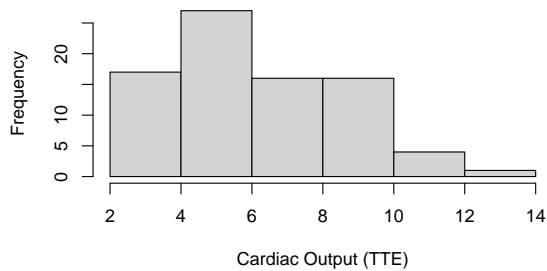
Parte 2: Análisis principales Gasto Cardíaco

J. Mancilla Galindo, E. Garza Santiago, L.A. Fernández Urrutia
A. Kammar-García

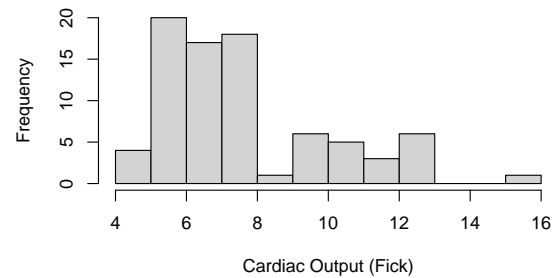
2024-12-10

Cardiac Output

The distribution of cardiac output values for both methods is skewed as shown bellow.



(a) TTE

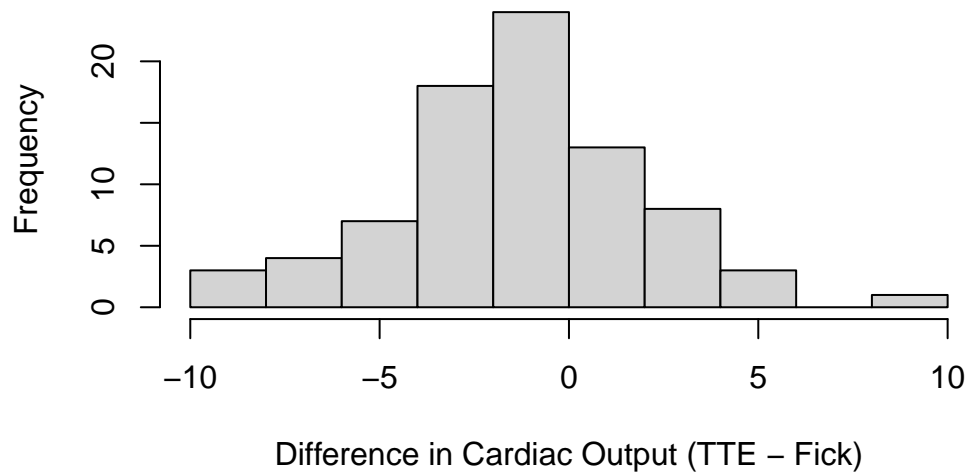


(a) Fick

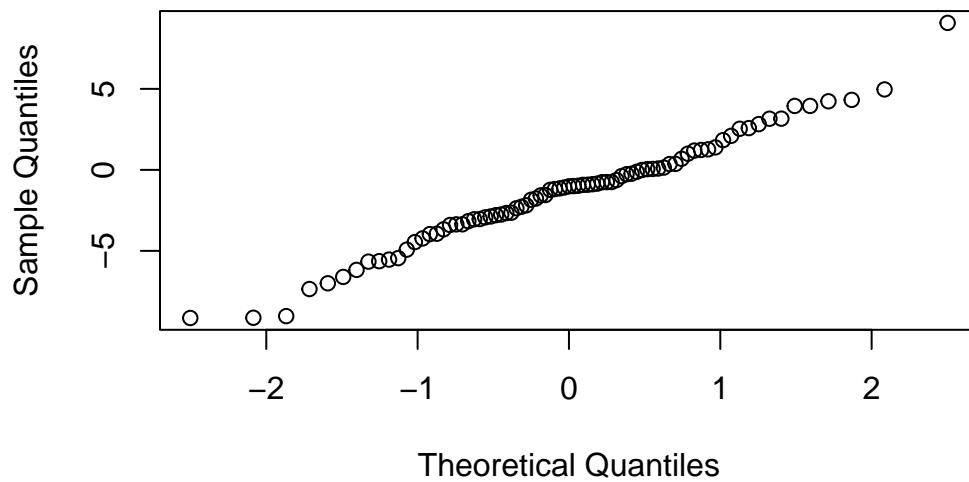
Thus, bootstrapping is used to calculate the mean with 95% CI:

The mean cardiac output for TTE is 6.26 L/min (95% CI: 5.73 to 6.81), and for Fick, 7.62 L/min (95% CI: 7.14 to 8.18).

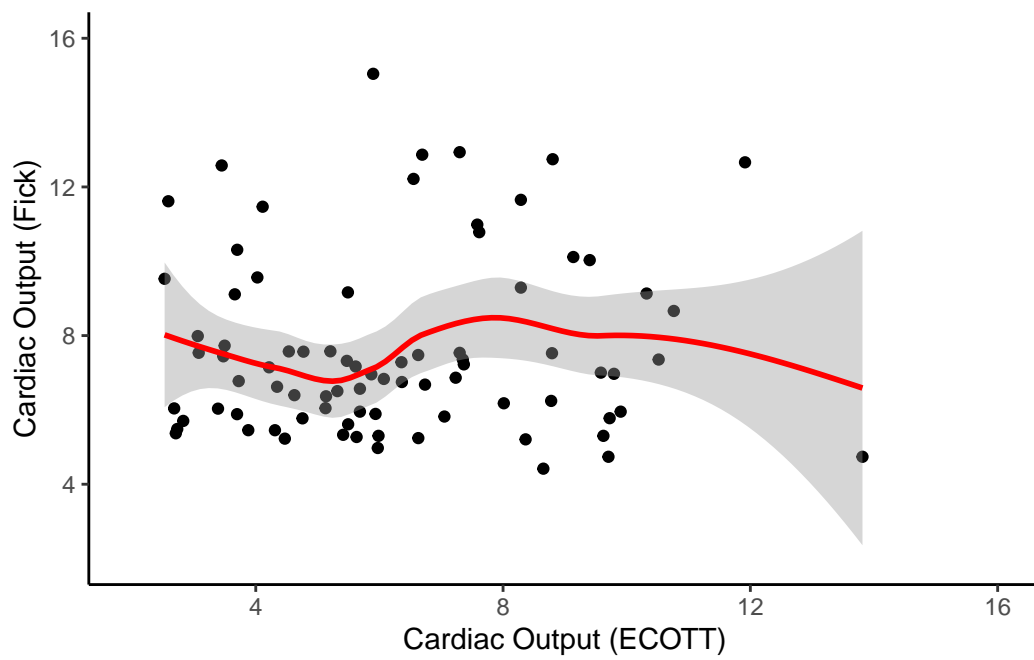
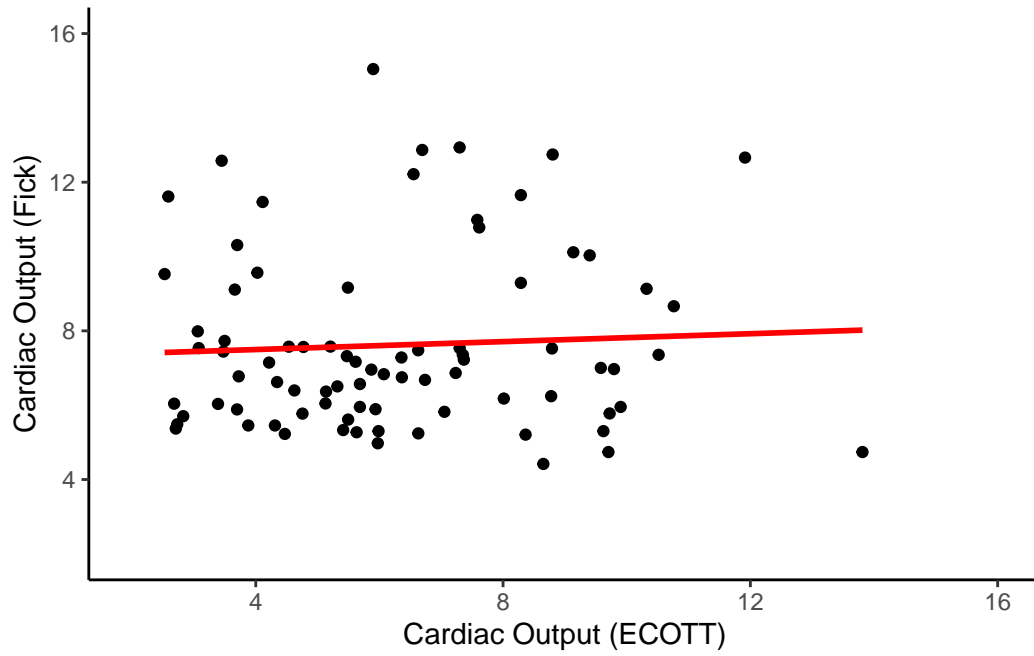
However, the distribution of differences between the two methods is approximately normal.



Normal Q-Q Plot



Assuming a linear and non-linear relationship



Examine if non-linear term is significantly better than linear term

Family: gaussian
Link function: identity

Formula:
cardiac_output_Fick ~ s(cardiac_output_TTE)

Parametric coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	7.6173	0.2674	28.49	<2e-16 ***

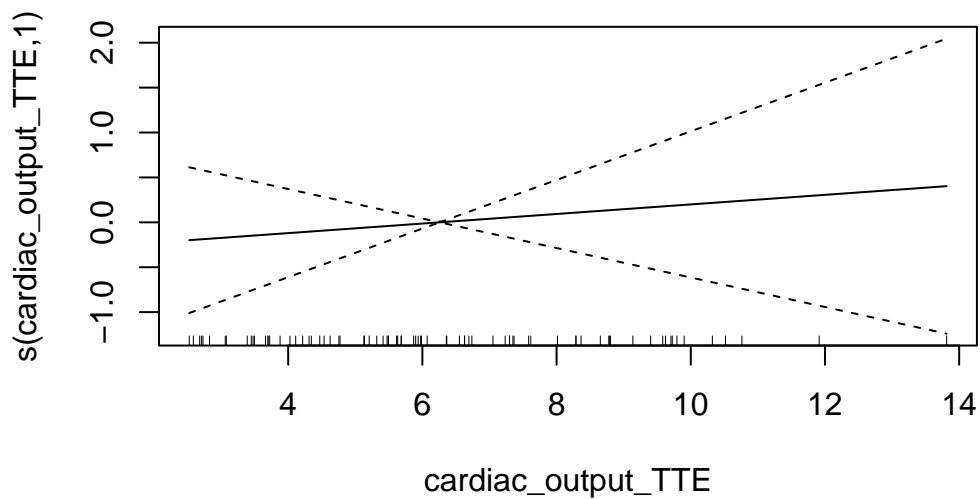
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Approximate significance of smooth terms:

	edf	Ref.df	F	p-value
s(cardiac_output_TTE)	1	1	0.24	0.626

R-sq.(adj) = -0.00959 Deviance explained = 0.303%

GCV = 5.9362 Scale est. = 5.7896 n = 81



Non-linear relationship is not significantly better than linear relationship. Thus, I will model as linear relationship.

Pearson correlation

Pearson's product-moment correlation

```
data: data$cardiac_output_TTE and data$cardiac_output_Fick
t = 0.48997, df = 79, p-value = 0.6255
alternative hypothesis: true correlation is not equal to 0
95 percent confidence interval:
 -0.1652936  0.2701451
sample estimates:
      cor
0.05504239
```

Linear regression

Call:

```
lm(formula = cardiac_output_TTE ~ cardiac_output_Fick, data = data)
```

Residuals:

Min	1Q	Median	3Q	Max
-3.8979	-1.8550	-0.3494	1.8022	7.7202

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	5.82344	0.92636	6.286	1.67e-08 ***
cardiac_output_Fick	0.05688	0.11608	0.490	0.626

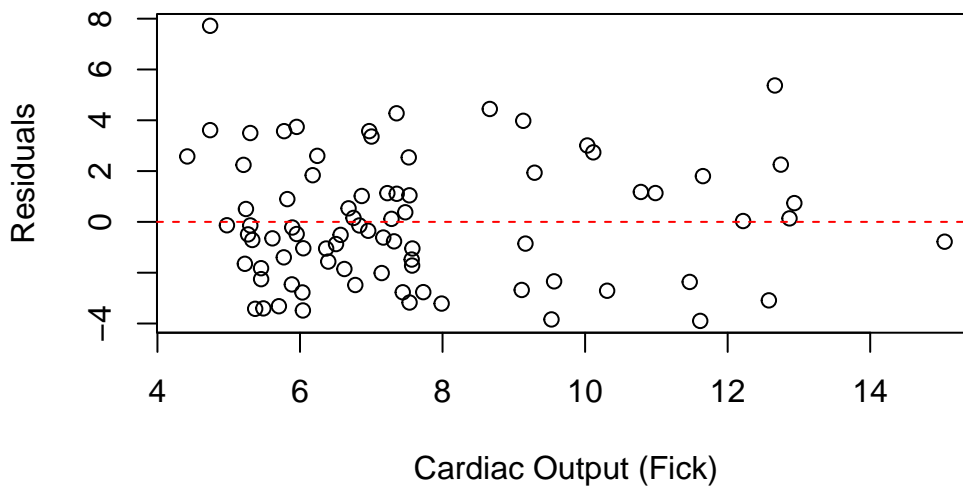
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 2.486 on 79 degrees of freedom

Multiple R-squared: 0.00303, Adjusted R-squared: -0.00959

F-statistic: 0.2401 on 1 and 79 DF, p-value: 0.6255

Residuals vs Cardiac Output (Fick)



Linear Mixed Effects Model

Linear mixed model fit by REML ['lmerMod']

Formula: cardiac_output_Fick ~ cardiac_output_TTE + (1 | ID) + (1 | time_point)

Data: data

REML criterion at convergence: 345.9

Scaled residuals:

Min	1Q	Median	3Q	Max
-1.70612	-0.35805	-0.09272	0.25930	2.18064

Random effects:

Groups	Name	Variance	Std.Dev.
ID	(Intercept)	3.170	1.780
time_point	(Intercept)	5.995	2.448
Residual		1.240	1.114

Number of obs: 81, groups: ID, 52; time_point, 9

Fixed effects:

	Estimate	Std. Error	t value
(Intercept)	9.13612	1.18262	7.725
cardiac_output_TTE	-0.01155	0.09538	-0.121

Correlation of Fixed Effects:

	(Intr)
crdc_tp_TTE	-0.570

Intraclass correlation coefficient (ICC)

Average Score Intraclass Correlation

Model: twoway

Type : agreement

Subjects = 81

Raters = 2

ICC(A,2) = 0.0917

F-Test, $H_0: r_0 = 0$; $H_1: r_0 > 0$

$F(80,79.6) = 1.12$, $p = 0.312$

95%-Confidence Interval for ICC Population Values:

$-0.325 < ICC < 0.39$

Coefficient of variation (CV) and coefficient of error (CE)

The following calculation is the coefficient of variation (CV) for the overall averaged measurements, expressed as percentage:

Fick CV: 31.44%

TTE CV: 39.55%

Because there are multiple measurements that are averaged to produce the mean cardiac output for TTE, we can calculate the coefficient of error (CE) as suggested by Cecconi, et al.¹ The following calculations reproduce the structure of the table in their review article:

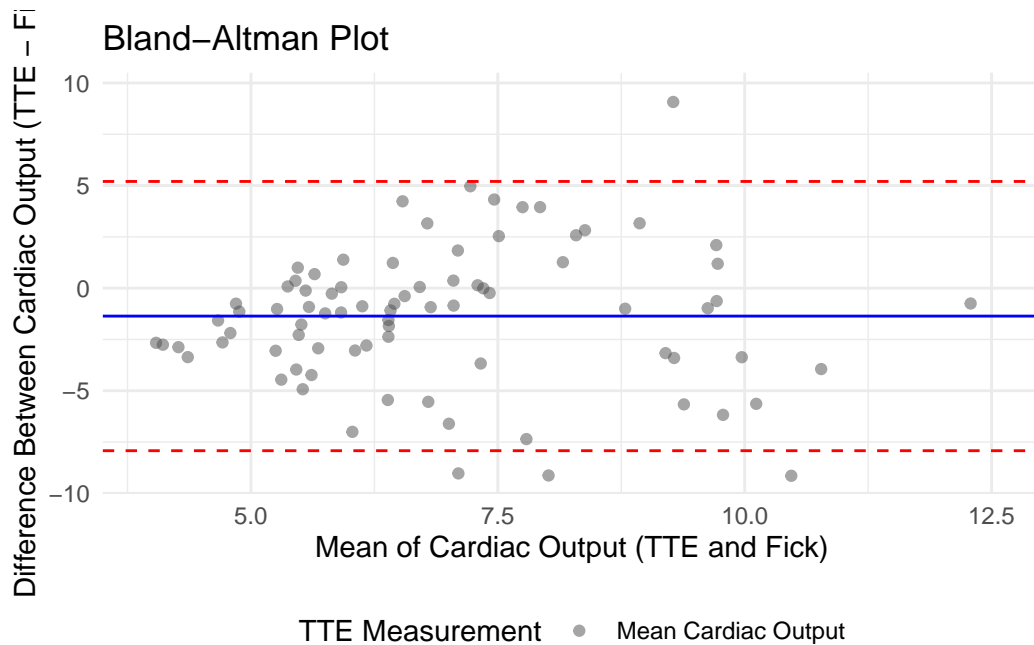
ID	CO 1 (TTE), L/min	CO 2 (TTE), L/min	CO 3 (TTE), L/min	Mean CO TTE, L/min	CV (%)	CE (%)	CO (Fick method), L/min
1	6.41	7.24	6.24	6.63	8.02	4.63	5.24
2	14.72	13.11	13.61	13.81	5.96	3.44	4.74
2	9.93	9.87	9.31	9.70	3.52	2.03	4.74
3	3.37	3.55	3.43	3.45	2.74	1.58	12.58
3	3.95	4.10	4.03	4.03	1.83	1.05	9.57
4	2.69	2.81	2.61	2.71	3.74	2.16	5.37

CE: Coefficient of Error; CV: Coefficient of Variation; TTE: Transthoracic Echocardiography.

The mean CV of TTE for the repeated measurements per patient was 7.3% (95% CI: 6.2 to 8.7) and the CE was 4.2% (95% CI: 3.6 to 5), corresponding to a precision of 8.4% (95% CI: 7.2 to 10.1).

Bland-Altman Plot

Bland Altman-single measure



Systematic bias (Paired t-test)

Paired t-test

```
data: data$cardiac_output_TTE and data$cardiac_output_Fick
t = -3.6582, df = 80, p-value = 0.0004534
alternative hypothesis: true mean difference is not equal to 0
95 percent confidence interval:
 -2.1007836 -0.6204367
sample estimates:
mean difference
 -1.36061
```

Bland Altman-repeated measures (random effects for between-subject variance)

Linear mixed model fit by REML ['lmerMod']

Formula: differences ~ 1 + (1 | ID)

Data: data

REML criterion at convergence: 409.7

Scaled residuals:

	Min	1Q	Median	3Q	Max
	-2.33206	-0.35646	-0.01357	0.29854	2.12253

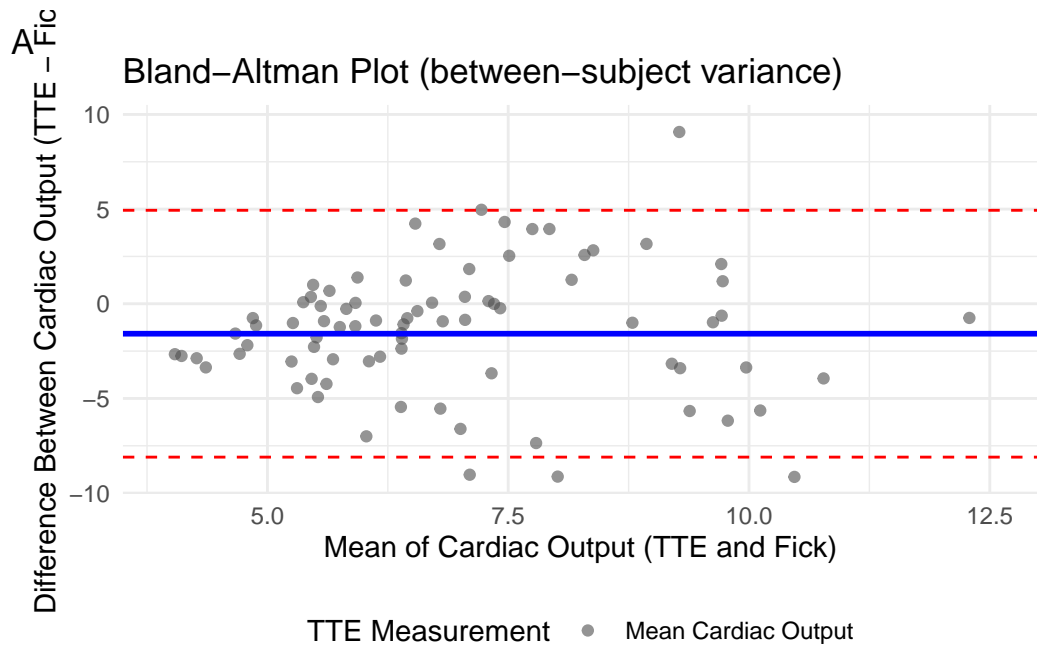
Random effects:

Groups	Name	Variance	Std.Dev.
ID	(Intercept)	6.149	2.480
Residual		4.909	2.216

Number of obs: 81, groups: ID, 52

Fixed effects:

	Estimate	Std. Error	t value
(Intercept)	-1.5811	0.4384	-3.606



Bland Altman-repeated measures (random effects for between-subject variance and within-subject variance)

Linear mixed model fit by REML ['lmerMod']

Formula:

differences ~ 1 + (1 | ID) + (1 | ID:time_point) + (1 | TTE_measurement)

Data: data_long

REML criterion at convergence: 765.2

Scaled residuals:

Min	1Q	Median	3Q	Max
-2.87065	-0.39608	-0.02016	0.39620	2.35851

Random effects:

Groups	Name	Variance	Std.Dev.
ID:time_point	(Intercept)	4.8084	2.1928
ID	(Intercept)	6.1494	2.4798
TTE_measurement	(Intercept)	0.0000	0.0000
Residual		0.3033	0.5508

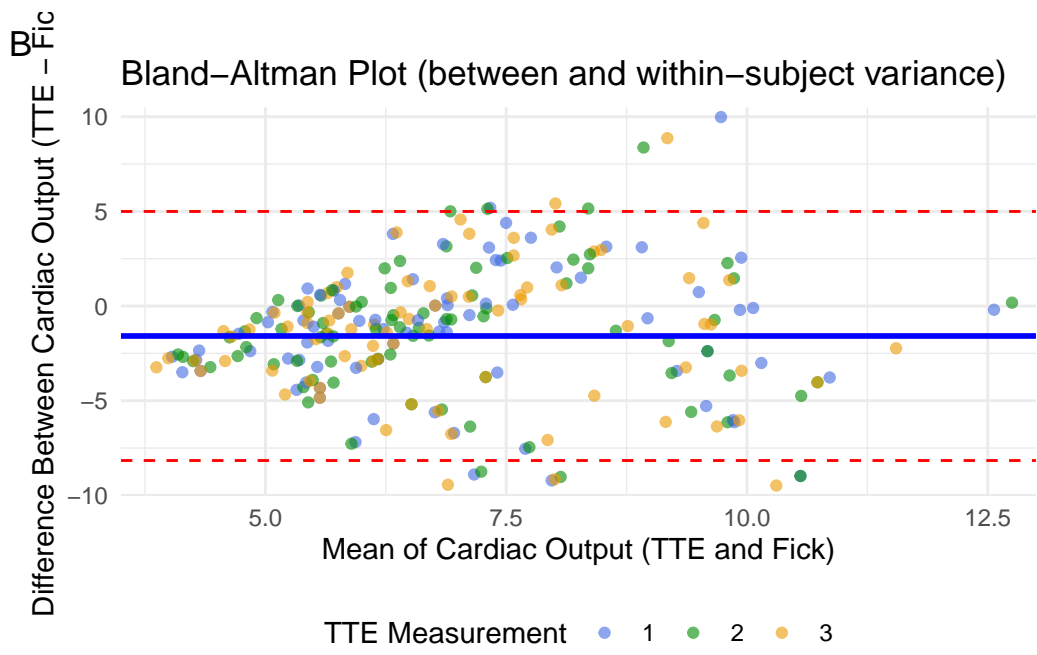
Number of obs: 243, groups: ID:time_point, 81; ID, 52; TTE_measurement, 3

Fixed effects:

	Estimate	Std. Error	t value
(Intercept)	-1.5811	0.4384	-3.606

optimizer (nloptwrap) convergence code: 0 (OK)
boundary (singular) fit: see help('isSingular')

There was singularity in the prior model including a random effect for within-subject TTE measurements. Because this term is conceptually important to take into account the nested structure of the data, we will keep it in the model. Other alternatives would be to include it as a fixed effect, but this would not necessarily represent a meaningful variable to model.



Mean absolute difference (MAD)

The MAD for cardiac output is 2.8 (95% CI: 2 to 3.5).

Mean absolute percentage error (MAPE) and precision of Fick method

Precisionb (point estimate): 56.84 %

Precisionb (95% CI): 44.34 - 74.22 %

Summary

The mean cardiac output with TTE was 6.26 (95% CI: 5.73 to 6.81) L/min, and 7.62 L/min (95% CI: 7.14 to 8.18) for the Fick method. The correlation between the two methods was $\rho = 0.06$ (95% CI: -0.17 to 0.27, $p=0.626$). In a linear mixed model with random patient slopes, there was a change in Fick CO of -0.01 (95% CI: -0.2 to 0.18) L/min for each unit change in mean TTE CO. The ICC between TCE and Fick CO 0.09 (95% CI: -0.32 to 0.39).

The mean absolute difference in CO between TTE and Fick was 2.76 (95% CI: 2 to 3.5) L/min. The coefficient of variation for an individual measurement of TTE was 39.5% and 31.44% for Fick. The mean CV of TTE for the repeated measurements per patient was 7.3% (95% CI: 6.2 to 8.7) and the CE was 4.2% (95% CI: 3.6 to 5), corresponding to a precision of 8.4% (95% CI: 7.2 to 10.1). The MAPE of the Fick method compared to TTE was 57.5% (95% CI: 45.5 - 74.6). The precision of the Fick method was 56.84% (95% CI: 44.34 to 74.22). The LSC was 11.9% (95% CI: 10.1 to 14.3) for TTE and 80.4% (95% CI: 62.7 to 105) for the Fick method.

Figure 1 shows the Bland-Altman plot for the repeated measures model with random effects for between-subject variance (Figure1A) and within-subject variance (Figure1B). The mean difference (systematic bias) between TTE and Fick CO was -1.58 (95% CI: -2.44 to -0.72, $p = 0$) L/min, with 95% limits of agreement of -8.16 to 5 L/min.

References

1. Cecconi M, Rhodes A, Poloniecki J, Della Rocca G, Grounds RM. Bench-to-bedside review: The importance of the precision of the reference technique in method comparison studies – with specific reference to the measurement of cardiac output. *Critical Care*. 2009;13(1):201. doi:[10.1186/cc7129](https://doi.org/10.1186/cc7129)

Package References

- Angelo Canty, B. D. Ripley (2024). *boot: Bootstrap R (S-Plus) Functions*. R package version 1.3-31. A. C. Davison, D. V. Hinkley (1997). *Bootstrap Methods and Their Applications*. Cambridge University Press, Cambridge. ISBN 0-521-57391-2, [doi:10.1017/CBO9780511802843](https://doi.org/10.1017/CBO9780511802843).
- Auguie B (2017). *gridExtra: Miscellaneous Functions for “Grid” Graphics*. R package version 2.3, <https://CRAN.R-project.org/package=gridExtra>.
- Bates D, Mächler M, Bolker B, Walker S (2015). “Fitting Linear Mixed-Effects Models Using lme4.” *Journal of Statistical Software*, 67(1), 1-48. doi:10.18637/jss.v067.i01 <https://doi.org/10.18637/jss.v067.i01>.
- Bates D, Maechler M, Jagan M (2024). *Matrix: Sparse and Dense Matrix Classes and Methods*. R package version 1.7-0, <https://CRAN.R-project.org/package=Matrix>.
- Berkelaar M (2024). *lpSolve: Interface to ‘Lp_solve’ v. 5.5 to Solve Linear/Integer Programs*. R package version 5.6.22, <https://CRAN.R-project.org/package=lpSolve>.
- Gamer M, Lemon J, puspendra.pusp22@gmail.com IFPS (2019). *irr: Various Coefficients of Interrater Reliability and Agreement*. R package version 0.84.1, <https://CRAN.R-project.org/package=irr>.
- Grolemund G, Wickham H (2011). “Dates and Times Made Easy with lubridate.” *Journal of Statistical Software*, 40(3), 1-25. <https://www.jstatsoft.org/v40/i03/>.
- Iannone R, Cheng J, Schloerke B, Hughes E, Lauer A, Seo J, Brevoort K, Roy O (2024). *gt: Easily Create Presentation-Ready Display Tables*. R package version 0.11.0, <https://CRAN.R-project.org/package=gt>.
- Makowski D, Lüdtke D, Patil I, Thériault R, Ben-Shachar M, Wiernik B (2023). “Automated Results Reporting as a Practical Tool to Improve Reproducibility and Methodological Best Practices Adoption.” *CRAN*. <https://easystats.github.io/report/>.
- Müller K, Wickham H (2023). *tibble: Simple Data Frames*. R package version 3.2.1, <https://CRAN.R-project.org/package=tibble>.
- Pinheiro J, Bates D, R Core Team (2023). *nlme: Linear and Nonlinear Mixed Effects Models*. R package version 3.1-164, <https://CRAN.R-project.org/package=nlme>. Pinheiro JC, Bates DM (2000). *Mixed-Effects Models in S and S-PLUS*. Springer, New York. doi:10.1007/b98882 <https://doi.org/10.1007/b98882>.
- R Core Team (2024). *R: A Language and Environment for Statistical Computing*. R Foundation for Statistical Computing, Vienna, Austria. <https://www.R-project.org/>.
- Rich B (2023). *table1: Tables of Descriptive Statistics in HTML*. R package version 1.4.3, <https://CRAN.R-project.org/package=table1>.
- Rinker TW, Kurkiewicz D (2018). *pacman: Package Management for R*. version 0.5.0, <http://github.com/trinker/pacman>.
- Wickham H (2016). *ggplot2: Elegant Graphics for Data Analysis*. Springer-Verlag New York. ISBN 978-3-319-24277-4, <https://ggplot2.tidyverse.org>.
- Wickham H (2023). *forcats: Tools for Working with Categorical Variables (Factors)*. R package version 1.0.0, <https://CRAN.R-project.org/package=forcats>.

- Wickham H (2023). *stringr: Simple, Consistent Wrappers for Common String Operations*. R package version 1.5.1, <https://CRAN.R-project.org/package=stringr>.
- Wickham H, Averick M, Bryan J, Chang W, McGowan LD, François R, Golemund G, Hayes A, Henry L, Hester J, Kuhn M, Pedersen TL, Miller E, Bache SM, Müller K, Ooms J, Robinson D, Seidel DP, Spinu V, Takahashi K, Vaughan D, Wilke C, Woo K, Yutani H (2019). “Welcome to the tidyverse.” *Journal of Open Source Software*, 4(43), 1686. doi:10.21105/joss.01686 <https://doi.org/10.21105/joss.01686>.
- Wickham H, François R, Henry L, Müller K, Vaughan D (2023). *dplyr: A Grammar of Data Manipulation*. R package version 1.1.4, <https://CRAN.R-project.org/package=dplyr>.
- Wickham H, Henry L (2023). *purrr: Functional Programming Tools*. R package version 1.0.2, <https://CRAN.R-project.org/package=purrr>.
- Wickham H, Hester J, Bryan J (2024). *readr: Read Rectangular Text Data*. R package version 2.1.5, <https://CRAN.R-project.org/package=readr>.
- Wickham H, Vaughan D, Girlich M (2024). *tidyr: Tidy Messy Data*. R package version 1.3.1, <https://CRAN.R-project.org/package=tidyr>.
- Wood SN (2011). “Fast stable restricted maximum likelihood and marginal likelihood estimation of semiparametric generalized linear models.” *Journal of the Royal Statistical Society (B)*, 73(1), 3-36. Wood S, N., Pya, S”afken B (2016). “Smoothing parameter and model selection for general smooth models (with discussion).” *Journal of the American Statistical Association*, 111, 1548-1575. Wood SN (2004). “Stable and efficient multiple smoothing parameter estimation for generalized additive models.” *Journal of the American Statistical Association*, 99(467), 673-686. Wood S (2017). *Generalized Additive Models: An Introduction with R*, 2 edition. Chapman and Hall/CRC. Wood SN (2003). “Thin-plate regression splines.” *Journal of the Royal Statistical Society (B)*, 65(1), 95-114.