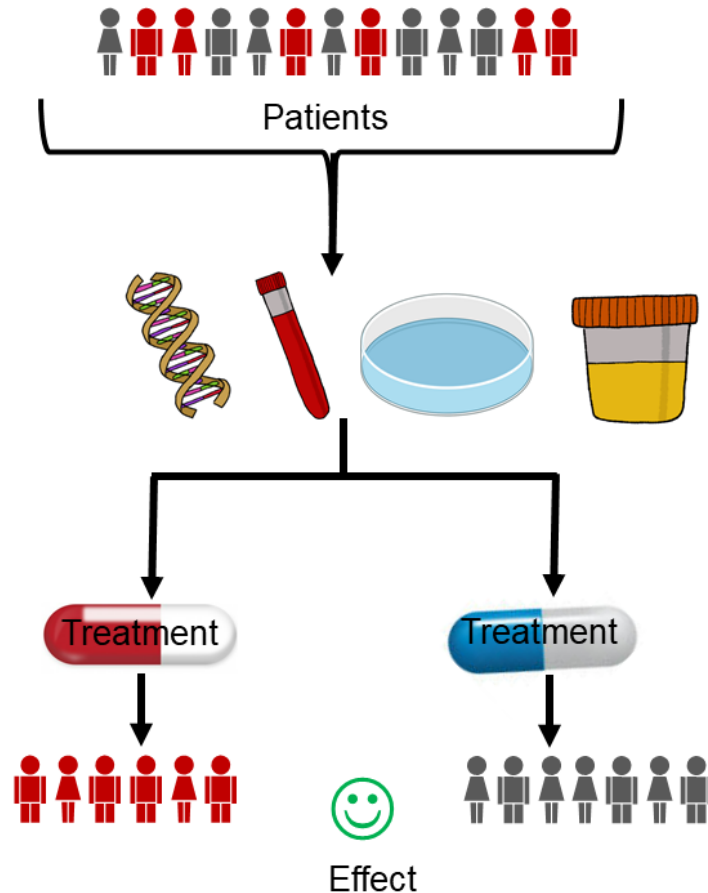


# Estimands in personalized medicine



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# Personalized medicine

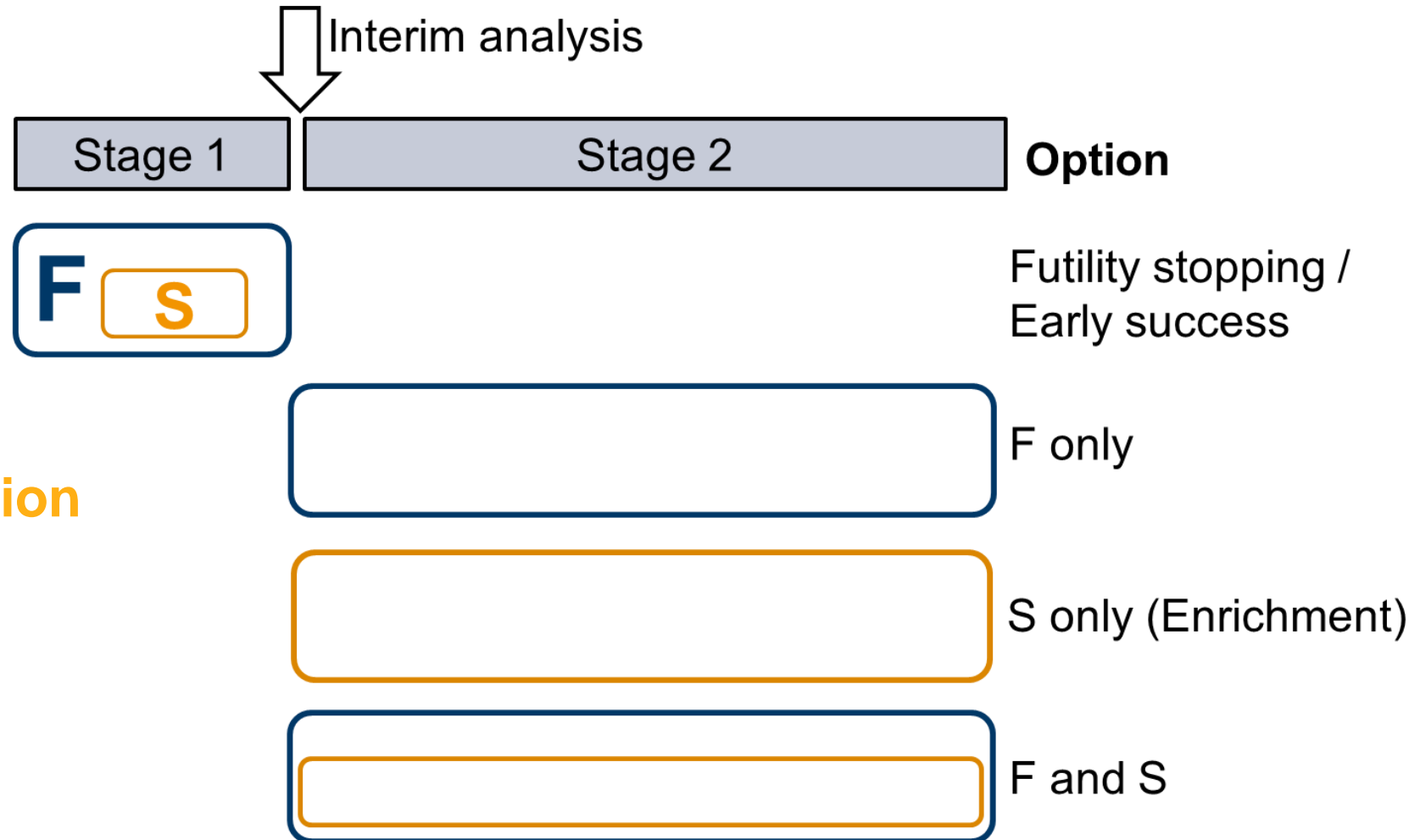


- Right drug for the right patient at the right time
- Choice of treatment
  - Accounting for disease and patient characteristics
  - Greatest benefit and least safety concerns (compared to alternatives)
- Stratification of patient population which may **differ in the efficacy** (or safety) of a specific treatment
  - Subgroups defined by **predictive** biomarkers

# Clinical trial designs for personalized medicine

- Assuming that a treatment works differently in different subgroups of patients
  - Randomized control trials **enrolling all patients not necessarily most efficient approach**
  - Enrichment designs (Temple, 2010)
    - Recruit only patients likely to benefit, e.g. biomarker-positive patients
    - **Risk of missing out** on subgroups that could benefit from treatment
- Data should be generated in both BM+ and BM- patients from a regulatory and public health perspective
  - Evidence on treatment being beneficial in biomarker-positive (BM+) patients, but uncertainty regarding benefit in biomarker-negative (BM-) patients
  - **Complex innovative trial designs for personalized medicine**, e.g. adaptive designs

# Adaptive enrichment design



**BM+ sub-population**  
**Full population**

# Estimands for adaptations of study population

- **Dual objective** of assessing efficacy in both BM+ patients and full population
  - Definition of **two separate estimands** in BM+ and in full population (Collignon et al, 2022)
  - **Estimands differ** in definition of **population**
  - Other attributes could but do not necessarily have to be the same
- Information regarding **efficacy in BM- needed** for informing the decision of approval and reimbursement in either the full population or the BM+ subgroup only (Collignon et al, 2022)
  - **Estimand for BM-** may be important
  - Differs from estimands of full and BM+ population by the "population" attribute
- **Discontinuation** of the BM- subpopulation at interim
  - Continuation of trial with a **single estimand**

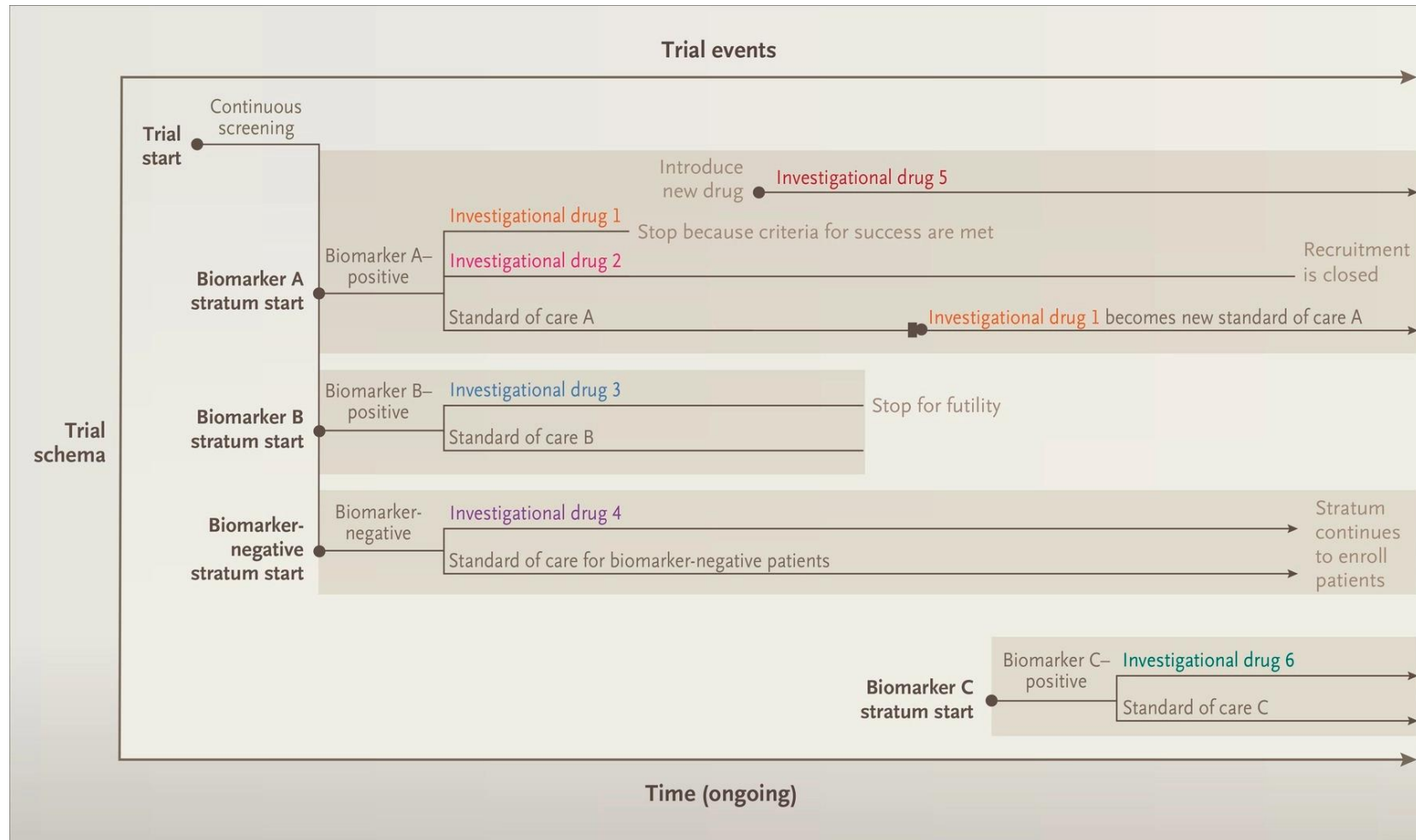
# Other attributes of estimands framework

- Sometimes the **primary endpoint only available after long-term follow-up**
  - Subgroup selection needs to be based **on early outcome data** (Friede et al ,2011; Kunz et al, 2012)
    - E.g. PFS and OS in oncology (Jenkins et al, 2011)
  - Different summary effect measure might be needed
  - **Estimands** at interim analysis and final analysis **might differ** in “**variable**” and “**population-level summary**” attributes
- Meaning of **intercurrent events** might be different at interim and final analysis
  - E.g. Treatment discontinuation less likely at early interim analysis
- Some **intercurrent events** might only occur in one of the subgroups, i.e. BM+ or BM-
  - E.g. Different safety and/or efficacy profile in subgroups leading to treatment discontinuation due to adverse events/ lack of efficacy in just one of the subgroups

# Other adaptations

- **Optimal cut-off** could be adaptively chosen in a trial
  - Biomarkers often measured on a continuous scale
  - Cut-off selection needed for defining BM+ and BM- subgroups
  - Pre-specification of a continuum of subgroups
    - **Continuum of pre-specified estimands** (Collignon et al, 2022)
    - Only one of these estimands will be chosen
- **Data-driven subgroup selection** at interim analysis, e.g. Johnston (2021)
  - Pre-specification of subgroup identification method
  - No pre-specification of subgroups, i.e. biomarkers and corresponding cut-offs
  - “Population” attribute cannot be described initially

# Platform trial



Woodcock and LaVange (2017)



# Estimands in platform trials

- Adding a new treatment arm to a platform trial
  - Adding another objective (comparing the new treatment vs. control)
  - Adding another estimand
- Prespecification of estimand not necessarily at start of platform trial
  - But before adding the new treatment arm (Collignon et al, 2022)
- Interim analyses in platform trials for dropping ineffective treatment arms
  - Trial continuation with remaining estimands

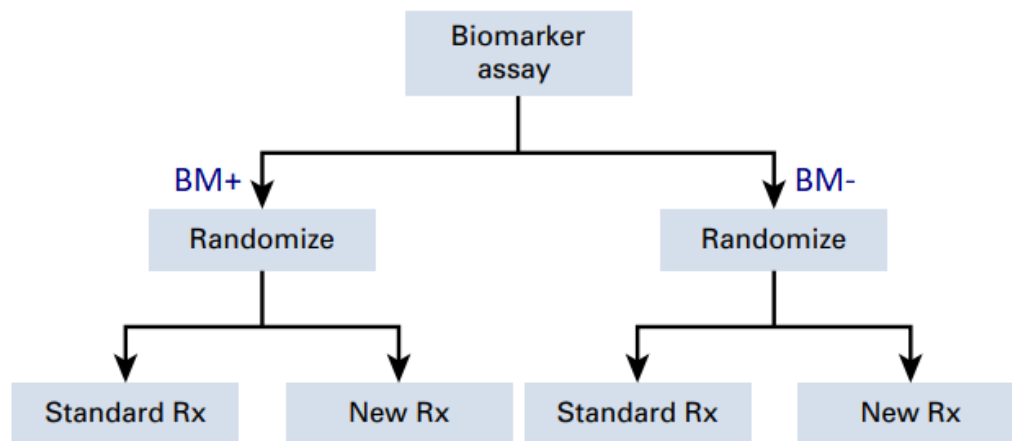
# Modification of treatment

- New evidence during a running trial can lead to a **modification of the control arm**
  - E.g. Approval of a **new treatment replacing** the original **standard of care**
  - **Two stages in the trial**: One with control treatment 1 and one with control treatment 2
- Comparable situation in platform trials when new treatment is added as data accrue
- Patients recruited to control prior to the addition of the new treatment/ prior to new control treatment are **nonconcurrent**
  - Discussion on whether or not including nonconcurrent patients for estimation in literature, e.g. Lee and Wason (2020), Lee et al (2021)
  - If main interest is comparison of treatments with a control regardless of any changes to the standard of care throughout the trial
    - Treatment attribute of estimands would remain aspecific: state-of-the-art control therapy
    - No update to estimands needed

# Estimation

- Selection of a subgroup at the interim analysis (Kimani et al, 2015)
  - Estimates are expected to be biased
  - Adjustment of confidence intervals and hypothesis test needed to control the error probabilities for the selected estimand
- For platform trials adding treatments during the trial
  - Proposals of borrowing information from nonconcurrent controls
  - Less bias by borrowing information within a trial than from external trials (Burger et al, 2021)
    - Many standardized aspects of trial less likely to cause bias
  - Adjustment of time trends needed when nonconcurrent controls are used
    - Otherwise, risk of bias (Lee and Wason, 2020; Dodd et al, 2021)

# Collapsibility



## Biomarker-stratified design

- Common practice for estimating **efficacy in full population**: Combining the strata and comparing the drug to the control in a meta-analytic approach (Okwuokenye, 2019)
- Estimates should be collapsible over subgroups
- Estimate of full population is collapsible if it is a weighted average of estimates in the subgroups, i.e. BM+ and BM-
- Odds ratios (OR) and hazard ratios (HR) are non-collapsible (Didelez et al, 2021)
- OR and HR can make a prognostic biomarker appear predictive even in RCTs (Liu et al 2022)

# Discussion

- Recent review on estimands and complex innovative designs including master protocols, adaptive designs by Collignon et al (2022)
  - Estimand framework is applicable to complex innovative designs which are also used for personalized medicine
- All five attributes could be affected using adaptive designs and master protocols for personalized medicine purposes
- Development of new adaptive designs using data-driven subgroup identification
  - Lack of experience in practice of such designs
  - Lack of experience on the applicability of the estimand framework

# References I

1. Burger, H.U. et al. (2021) The use of external controls: to what extent can it currently be recommended? *Pharm. Stat.* 20, 1002–1016.
2. Collignon, O., Schiel, A., Burman, C.-F., Rufibach, K., Posch, M. and Bretz, F. (2022), Estimands and Complex Innovative Designs. *Clin Pharmacol Ther*, 112: 1183-1190. <https://doi.org/10.1002/cpt.2575>
3. Didelez, V, Stensrud, MJ. On the logic of collapsibility for causal effect measures. *Biometrical Journal*. 2022; 64: 235–242. <https://doi.org/10.1002/bimj.202000305>
4. Dodd, L.E., Freidlin, B. & Korn, E.L. (2021) Platform trials – beware the noncomparable control group. *N. Engl. J. Med.* 384, 1572–1573.
5. Friede, T., Parsons, N., Stallard, N., Todd, S., Valdés-Márquez, E., Chataway, J., & Nicholas, R. (2011). Designing a seamless phase II/III clinical trial using early outcomes for treatment selection: An application in multiple sclerosis.
6. Friede, T., Parsons, N. and Stallard, N. (2012), A conditional error function approach for subgroup selection in adaptive clinical trials. *Statist. Med.*, 31: 4309-4320. <https://doi.org/10.1002/sim.5541>
7. Jenkins, M., Stone, A. and Jennison, C. (2011), An adaptive seamless phase II/III design for oncology trials with subpopulation selection using correlated survival endpoints. *Pharmaceut. Statist.*, 10: 347-356. <https://doi.org/10.1002/pst.472>

## References II

7. Johnston, S. E., Lipkovich, I., Dmitrienko, A., & Zhao, Y. D. (2022). A two-stage adaptive clinical trial design with data-driven subgroup identification at interim analysis. *Pharmaceutical statistics*, 21(5), 1090–1108.
8. Kimani, P. K., Todd, S., & Stallard, N. (2015). Estimation after subpopulation selection in adaptive seamless trials. *Statistics in medicine*, 34(18), 2581–2601. <https://doi.org/10.1002/sim.6506>
9. Kunz, C. U., Friede, T., Parsons, N., Todd, S., & Stallard, N. (2014). Data-driven treatment selection for seamless phase II/III trials incorporating early-outcome data. *Pharmaceutical Statistics*, 13, 238– 246.
10. Lee, K.M., Wason, J. (2020). Including non-concurrent control patients in the analysis of platform trials: is it worth it?. *BMC Med Res Methodol* 20, 165. <https://doi.org/10.1186/s12874-020-01043-6>
11. Lee, K.M., Brown, L.C., Jaki, T., Stallard, N. & Wason, J. (2021) Statistical consideration when adding new arms to ongoing clinical trials: the potentials and the caveats. *Trials* 22, 203.
12. Liu, Y, Wang, B, Yang, M, Hui, J, Xu, H, Kil, S, Hsu, JC. (2022) Correct and logical causal inference for binary and time-to-event outcomes in randomized controlled trials. *Biometrical Journal*; 64: 198-224.

## References III

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13. Okwuokenye, M. and Peace, KE. (2019), Adaptive Design and the Estimand Framework. *Annal Biostat & Biomed Appl*, 1(5). <https://doi.org/10.33552/ABBA.2019.01.000524>
14. Temple, R. (2010), Enrichment of Clinical Study Populations. *Clinical Pharmacology & Therapeutics*, 88: 774-778. <https://doi.org/10.1038/clpt.2010.233>
15. Woodcock, J., & LaVange, L. M. (2017). Master Protocols to Study Multiple Therapies, Multiple Diseases, or Both. *The New England journal of medicine*, 377(1), 62–70. <https://doi.org/10.1056/NEJMra1510062>