Headcount for Hashbrowns Biostatistics 699 Project #4 Interim Presentation

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April 4, 2019

Design Question 2

Background

Who are we trying to help?

- Some patients with cancers of the blood are receiving an alloHSCT from a genetically-matched donor because other forms of treatment are not working
- Among those who receive an alloHSCT, 40-50% experience an immune reaction called acute graft-versus-host disease (aGVHD)

What is aGVHD?

- GVHD occurs when the T-cells of the graft fight the remaining immune system of the host
- It is classified as acute if it occurs within 100 days
- One of the leading causes of early death in HSCT recipients with few approaches to prevent or treat

Background

What can be done?

- We know that this type of treatment alters the intestinal microbiome
- We hope that treatments that maintain health in the intestine reduce the incidence of aGVHD
- One potential treatment is an increased consumption of potato starch

What is our job?

Design a one-arm clinical trial in which all participants ingest
 40g of starch everyday for up to 100 days

Research Question 1

What is the sample size needed to detect a reduction of the 50% historical rate of aGVHD to each of the following

- 40% (small potato starch effect)
- 35% (medium potato starch effect)
- 30% (large potato starch effect)

with 80% power and 5% type I error?

Pertinent Information

Distribution of time-to-onset of aGVHD

1st Quartile: 7 days

Median: 28 days

3rd Quartile: 56 days

Competing Risks

 15% will experience some form of competing risk unrelated to the treatment and will not be observable for aGVHD

Patient Noncompliance

20% will discontinue taking starch unrelated to aGVHD

- Simulate survival data to replicate each treatment effect
- With this data, plot the Kaplan-Meier curve and determine if it is significantly different from 50% survival at Day 100 by looking at the 95% confidence interval
- Repeat Steps 1 and 2 many times and the proportion of significant findings is the power
 - Repeated Steps 1 and 2 ten thousand times
 - Repeated Steps 1-3 for different sample sizes (10 to 300, by 5)

Simulate Survival Data

Sample from a Binomial(N, p) to determine those who get aGVHD during treatment

- N is the sample size
- p is one of three treatment effect sizes

Sample from a **Truncated Exponential Distribution** to determine the survival time of those who get aGVHD

- Truncated at 100 days to follow the definition of aGVHD
- Those who do not get aGVHD given an arbitrary survival time greater than 100 (e.g. 110 days)

The Truncated Exponential Distribution

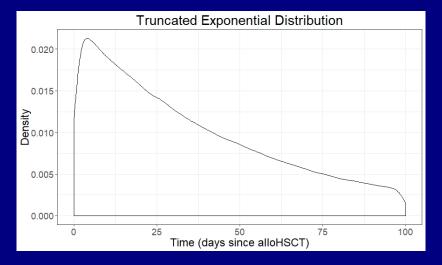


Figure 1: Sampling distribution of aGVHD times

The Truncated Exponential Distribution

$$F(x) = \frac{1 - \exp(-x/\lambda)}{1 - \exp(-100/\lambda)}$$

Solve for λ such that F(28) = 0.5 (Median is 28 days)

$$\lambda = 49.06266$$

The interquartile range doesn't perfectly match up, but it is close enough

- 1st Quartile: 12.03 days (Target: 7 days)
- 3rd Quartile: 51.83 days (Target: 56 days)

Assumption: The distribution of survival times stays the same for any effect size

##		Time	aGVHD
##	1	110.00	0
##	2	13.95	1
##	3	110.00	0
##	4	67.83	1
##	5	110.00	0
##	6	21.44	1
##	7	32.43	1
##	8	110.00	0
##	9	18.77	1
##	10	110.00	0

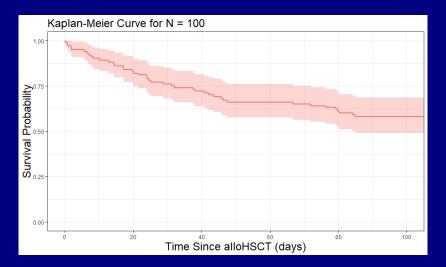


Figure 2: No competing risks and full patient compliance

Results

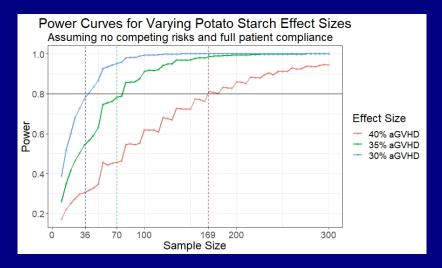


Figure 3: 10,000 Simulations

Adjusting for competing risks

- Independently sample from a Binomial(N, p = 0.15) to determine those who experience a competing risk
- The time-to-onset of the competing risk is sampled from a Uniform(0, 100)
- These people are censored, but provide information for as long as they were treated before the competing risk

Adjusting for noncompliance

- Independently sample from a Binomial(N, p = 0.20) to determine those who discontinue taking starch
- The time-to-noncompliance is sampled from a Uniform(0, 100)
- These people are censored, but provide information for as long as they were actually taking the treatment

##		Time	aGVHD	Time_CR	CR	Time_NC	NC	Time_E	Event
##	1	25.77	1	35.72	1	48.55	1	25.77	1
##	2	39.53	1	110.00	0	39.36	1	39.36	0
##	3	110.00	0	110.00	0	44.63	1	44.63	0
##	4	110.00	0	110.00	0	110.00	0	110.00	0
##	5	110.00	0	110.00	0	110.00	0	110.00	0
##	6	54.71	1	110.00	0	110.00	0	54.71	1
##	7	110.00	0	110.00	0	110.00	0	110.00	0
##	8	23.04	1	110.00	0	110.00	0	23.04	1
##	9	42.19	1	110.00	0	110.00	0	42.19	1
##	10	110.00	0	110.00	0	110.00	0	110.00	0

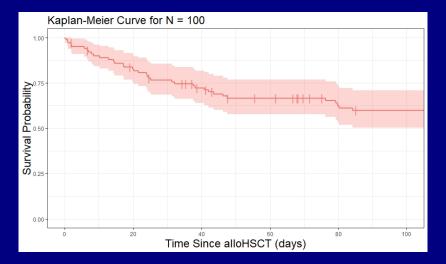


Figure 4: With 15% competing risks and 80% patient compliance

Results

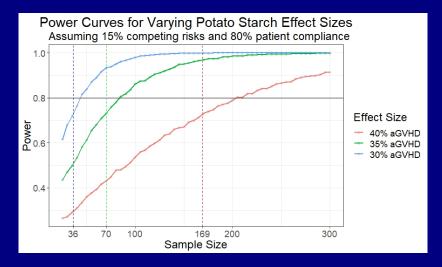


Figure 5: 10,000 Simulations

Research Question 2

How would you design the study to include one interim analysis of futility, and how does this affect the sample size you found previously?

to be determined . . .

Saw-Toothed Behavior

"For continuous random variables it is intuitively clear that for a given significance level and alternative hypothesis the power function increases monotonically as the sample size increases. One would expect the same to be true for discrete random variables, but such is not the case... The reason for this counter-intuitive result is that a significance level of 0.05, for example, cannot be obtained exactly for most sample sizes. So by a 0.05 significance level for a given sample size n we mean the largest level below 0.05. That level determines the critical value for the test statistic and the critical value determines the power at any given alternative. If n is such that the exact significance level is appreciably lower than 0.05 (say 0.01) and the power is 80% at a particular alternative, the power could easily drop below 80% for n + 1 and the same alternative hypothesis, since the exact significance level does not remain at 0.01 but may stay below but get closer to 0.05. This is particularly the case for small n."