

# CTN-0094 Demographics Analysis

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## Quarto

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## Running Code

When you click the **Render** button a document will be generated that includes both content and the output of embedded code. You can embed code like this:

```
1 + 1
```

```
[1] 2
```

You can add options to executable code like this

```
[1] 4
```

The `echo: false` option disables the printing of code (only output is displayed).

## Introduction

- Illicit opioid use is a major public health crisis associated with high overdose mortality (OpenAI, 2025).
- Medications for opioid use disorder (MOUD) improve retention and reduce mortality.
- The NIDA Clinical Trials Network (CTN) runs large, real-world clinical trials evaluating MOUD effectiveness.
- CTN project **CTN-0094** harmonizes data from projects **CTN-0027**, **CTN-0030**, and **CTN-0051** (public.ctn0094data, 2025).

- This report summarizes baseline demographics across the three randomized CTN studies.
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## Methods

### Software versions

- **R version:** R version 4.5.2 (2025-10-31)
- **public.ctn0094data:** 1.0.6
- **tidyverse:** 2.0.0
- **gtsummary:** 2.4.0

### Load & prepare data

```
“‘{r} library(tidyverse) library(public.ctn0094data) library(gtsummary) library(labelled) library(glue) library(scales)
```

### Load datasets

```
demographics <- public.ctn0094data::demographics randomization <- public.ctn0094data::randomization
```

### Track raw size

```
n_demo_raw <- nrow(demographics)
```

### Identify FIRST randomization only

```
first_rand <- randomization |> arrange(who, randomization) |> filter(randomization == 1)  
|> distinct(who, .keep_all = TRUE)  
randomized_ids <- first_rand |> select(who)
```

## **Missing data before filtering**

```
n_no_age <- demographics |> filter(is.na(age)) |> nrow() n_no_sex <- demographics |> filter(is.na(is_male)) |> nrow()
```

## **Create analysis dataset**

```
analysis <- demographics |> semi_join(randomized_ids, by = "who") |> filter(!is.na(age), !is.na(is_male)) |> mutate( project = case_when( project == 27 ~ "CTN-0027", project == 30 ~ "CTN-0030", project == 51 ~ "CTN-0051", TRUE ~ as.character(project) ), project = factor(project, levels = c("CTN-0027","CTN-0030","CTN-0051")), is_male_factor = factor(if_else(is_male == 1, "Male", "Not Male")) ) |> labelled::set_variable_labels( who = "Participant ID", age = "Age at Intake", is_male_factor = "Assigned Male Sex at Birth" )  
n_analysis <- nrow(analysis) n_excluded <- n_demo_raw - n_analysis
```

## **Theme**

```
ctn_theme <- theme_minimal(base_size = 14) + theme( plot.title = element_text(face = "bold"), axis.title = element_text(face = "bold") )
```

## **Sex proportion differences**

```
sex_by_project <- analysis |> group_by(project) |> summarise(prop_male = mean(is_male_factor == "Male"))  
max_diff_male <- max(sex_by_project$prop_male) - min(sex_by_project$prop_male)  
max_diff_male_fmt <- percent(max_diff_male, accuracy = 0.1)  
clinically_meaningful <- 0.05 is_meaningful <- max_diff_male >= clinically_meaningful
```

## **Chi-square test (no rstatix)**

```
chi_data <- table(analysis$project, analysis$is_male_factor) chi_test <- chisq.test(chi_data)  
p_chi <- chi_test$p.value p_chi_fmt <- scales::pvalue(p_chi)
```