Creating ShinyR Dashboards using Al

For Malaysian Healthcare Decision-making

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NHTA Pre-conference Workshop

29 July 2025 | 2:00 PM - 5:00 PM | PICC Malaysia









- Introduction
- Shiny Fundamentals
- 3 LLM-Powered Development
- 4 Healthcare Dashboard Lab
- Conclusion

Workshop Agenda

Agenda

- 2:00-2:15: Introduction to Shiny & Healthcare Dashboards
- 2:15-2:45: Shiny Fundamentals
- 2:45-3:15: LLM-Powered Development (ChatGPT/Grok/Deepseek)
- 3:15-3:30: Coffee Break
- 3:30-4:50: Healthcare Dashboard Lab
- 4:50-5:00: Q&A & Resources

Why Shiny for Healthcare?

Healthcare Applications

- Patient outcome tracking
- Resource allocation monitoring
- Epidemiological trend analysis
- Clinical trial reporting

Benefits

- Real-time data visualisation
- Interactive decision support
- Customisable for Malaysian healthcare needs

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Shiny App Anatomy

Core Components

- UI (User Interface): Controls layout, inputs (e.g., shinyWidgets dropdowns)
- Server: Handles data processing, reactivity
- Reactivity: Automatic UI updates with modular code (e.g., moduleServer)
- Modularity: Use reusable UI/server modules for scalability

Healthcare-Ready Components

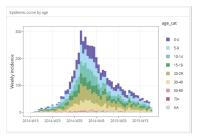
Inputs

- Date range selectors (e.g., shinyWidgets sliders)
- Patient group filters
- Clinical parameter sliders
- Facility selectors with input validation

Outputs

- Interactive epidemic curves (plotly/ggplot2)
- Patient outcome tables (DT)
- Resource utilisation charts
- Geographic heatmaps (leaflet)





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LLM Workflow for Shiny

Workflow Steps

- 1 Define clear requirements (objectives, data, visualisations, UI)
- @ Generate modular code with LLMs (e.g., "ShinyR module for plotly bar chart")
- Optimise prompts with R/Shiny versions, libraries, error handling
- 4 Validate, debug, and enhance interactivity (e.g., shinyWidgets, leaflet)

Effective Prompt Engineering

Good Prompt Structure

- "Create a ShinyR app (Shiny 1.8, R 4.4) using medicaldata::covid_testing:
- Sidebar with facility dropdown, date range input
- Plotly positivity rate trend chart
- DT table for patient demographics
- Add try-catch for missing data, input validation for positive numbers"

Key Elements

- Specify R/Shiny versions, libraries, dataset structure
- Define visualisations (e.g., plotly, leaflet)
- Include error handling, interactivity
- Use synthetic data for sensitive projects

Hands-on Time!

Steps

- Open RStudio and your preferred LLM (ChatGPT, Grok, Deepseek)
- Use prompt: "Create ShinyR app (Shiny 1.8) with NHANES data, plotly scatterplot of BMI vs. Height, age range slider, DT table, error handling for missing data."
- Copy, paste, and run the script in RStudio
- Compare outputs from different LLMs

Shiny App Code: Libraries and Data Preparation

```
# Load libraries
library(shiny)
library (NHANES)
library(plotly)
library(DT)
library(dplyr)
# Prepare NHANES data with error handling
data (NHANES)
nhanes_data <- tryCatch({
 NHANES %>%
    select(Age, BMI, Height) %>%
    filter(complete.cases(.)) # Remove rows with missing values
}, error = function(e) {
 message("Error, in, data, preparation:, e)
 return (NULL)
1)
```

Shiny App Code: UI Definition

```
# Define UI with modular structure
ui <- fluidPage(
 titlePanel("NHANES Dashboard"),
 sidebarLayout (
    sidebarPanel(
      sliderInput("age_range", "Select,Age,Range:",
                  min = min(nhanes_data$Age, na.rm = TRUE),
                  max = max(nhanes_data$Age, na.rm = TRUE),
                  value = c(min(nhanes_data$Age, na.rm = TRUE), max(nhanes_data$Age, na.rm = TRUE)),
                  step = 1),
      # Input validation
      tags$script("Shiny.addCustomMessageHandler('alert', ...function(message), [alert(message); ]);")
   ),
    mainPanel(
      plotlyOutput("scatterPlot").
      DTOutput ("dataTable")
```

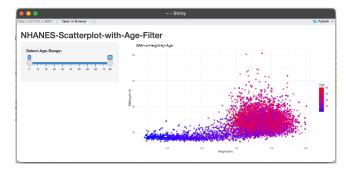
Shiny App Code: Server Logic and Execution

```
# Define server logic with modular reactivity
server <- function(input, output, session) {
 # Reactive data filtering
 filtered data <- reactive({
    req(nhanes_data)
    validate(need(input age_range[1] >= 0, "Age, must, be, positive"))
    nhanes data %>%
      filter(Age >= input$age_range[1] & Age <= input$age_range[2])
 3-)
 # Render plotly scatterplot
 output$scatterPlot <- renderPlotly({
    plot_ly(filtered_data(), x = "Height, y = "BMI, colour = "Age, type = "scatter", mode = "markers") %>%
      layout(title = "BMI, vs, Height, by, Age", xaxis = list(title = "Height, (cm)"), yaxis = list(title = "BMI
            (kg/m^2)"))
 1)
 # Render DT table
 output$dataTable <- renderDT({
    datatable(filtered_data(), options = list(pageLength = 10))
 1)
# Run the application
shinyApp(ui = ui, server = server)
```

NHANES Dashboard Example

Interactive Dashboard

Visualises BMI vs. Height with an age range filter and interactive DT table.



Explore live dashboards: Diabetes & Hypertension Insights, Malaysian News Hub

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Lab Exercise: COVID-19 Dashboard

Dataset and Requirements

Dataset: medicaldata::covid_testing

Requirements:

- Facility performance comparison (plotly bar chart)
- 2 Test result trends over time (plotly line chart)
- 3 Patient demographic filters (shinyWidgets dropdown)
- Outcome rate visualisation (leaflet map)

LLM Prompt Template

"Create a ShinyR dashboard for medicaldata::covid_testing:

- Sidebar with shinyWidgets facility dropdown, date range slider
- Plotly positivity rate trend chart, leaflet map for outcomes
- DT table for demographics, try-catch for missing data
- Add testthat unit tests for server logic."

Step-by-Step Implementation

Phase 1: Environment Setting

- Set up account on Posit (online RStudio): https://posit.cloud/
- Create a new project
- Prepare the coding canvas

Phase 2: Dashboard Development

- Specify the requirements for the dashboard
- Save the final code as "app.R" (for hosting/publishing purposes)

Phase 3: Hosting on a Website

- Set up account on shinyapps.io: https://www.shinyapps.io/
- 2 Create a token & copy the token
- Output
 Publish it (by using the token on another Posit coding canvas)!

Please find a detailed guideline HERE.



Relevance to Malaysian Healthcare

Key Requirements

- Patient Confidentiality: High relevance; supported by MMC guidelines for data privacy.
- Clinical Data Validation: Essential for EMR and trials, ensuring accuracy (study).
- Accessibility Compliance (WCAG): Relevant for digital inclusivity (global guide).
- Exportable Reports: Standard for decisions, e.g., MSOM.

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Key Takeaways

Summary

- Shiny enables rapid, interactive healthcare dashboards
- LLMs (ChatGPT, DeepSeek, or Grok) reduce coding barriers
- Ensure security (local LLMs, synthetic data) and compliance
- Stay updated with Shiny features, optimise, and test thoroughly

Resources

- Shiny Basics: https://shiny.posit.co/
- Healthcare Datasets:
 - ► NHANES: https://wwwn.cdc.gov/nchs/nhanes/
 - ► MIMIC: https://mimic.mit.edu/
 - ▶ medicaldata: https://higgi13425.github.io/medicaldata/
- LLM Platforms:
 - ► ChatGPT: https://chat.openai.com/
 - ► Grok: https://grok.com/
 - ► Deepseek: https://chat.deepseek.com/
- Advanced References:
 - ► EpiR Handbook:
 - https://epirhandbook.com/en/new_pages/shiny_basics.html
 - ► ShinyR dashboards: https://tilburgsciencehub.com/examples
 - ► R packages: https://cran.r-project.org/web/views/

Questions are welcomed!