Subject: Here is the way GPT thinks we should do it

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To generate a patient query entry for an LLM using FHIR data, particularly for the purpose of:

- 1. Computing an encoder vector (i.e., a numerical patient embedding for downstream modeling tasks)
- 2. Predicting risk of Treatment-Resistant Depression (TRD)

#### we need to:

- Convert FHIR-based structured EHR data into serialized input (usually Markdown or plain text with structure)
- Maintain temporal order, coding semantics, and clinical context
- Make the serialization LLM-friendly, meaning it's concise, flattened, and ready for language modeling or embedding



# Step-by-Step Framework

# 1. Define Scope of Relevant FHIR Resources

To capture information relevant to TRD prediction, include:

#### **FHIR Resource Key Fields**

Patient Age, sex, race/ethnicity, socioeconomic status

Condition Major depressive disorder (onset, course), comorbid anxiety, PTSD, etc. MedicationRequest Antidepressant class, dose, timing, number of trials, augmentation meds

Observation PHQ-9 scores, sleep/appetite/energy ratings, weight/BMI Encounter Visit frequency (outpatient/inpatient/ED), gaps in care

Procedure ECT, TMS, psychotherapy documented

DiagnosticReport Imaging (e.g., MRI), if relevant

AllergyIntolerance Treatment-limiting allergies (e.g., SSRI)

# 2. Convert FHIR to Text Prompt (Markdown or Plain Text)

Flatten into a **chronological**, **structured text block**. Here's an example:



# LLM Patient Input Example for TRD Risk Prediction

## Patient Demographics

- Age: 42
- Sex: Female
- Race: Black
- Insurance: Medicaid

# ## Diagnoses (Condition Resource)

- Major depressive disorder (ICD-10: F33.1) diagnosed 2021-03-01
- Generalized anxiety disorder (ICD-10: F41.1) diagnosed 2019-11-15
- Obesity (ICD-10: E66.9) diagnosed 2022-05-10

## Antidepressant Trials (MedicationRequest)

- 1. Sertraline 50–200mg daily started 2021-03-10, stopped 2021-06-15 (ineffective)
- 2. Bupropion XL 150mg → 300mg daily 2021-06-20 to 2021-12-10 (partial response)
- 3. Venlafaxine XR 75mg → 225mg 2022-01-05 to 2022-07-01 (poor tolerability)
- 4. Mirtazapine 15mg qhs 2022-08-15 to present (ongoing)

## ## Augmentation & Other Psychiatric Medications

- Aripiprazole 2mg added 2022-10-01 (partial benefit)
- Hydroxyzine PRN
- No lithium, ECT, or TMS

# ## PHQ-9 Scores (Observation)

- 2021-03-01: 18
- 2021-06-01: 15
- 2021-10-01: 17
- 2022-03-01: 19
- 2022-09-01: 16
- 2023-03-01: 18

# ## Clinical Events (Encounter + Procedure)

- 14 outpatient psych visits over 2 years
- No psychiatric hospitalizations
- No psychotherapy procedures documented

## ## Comorbidities

- Type 2 Diabetes
- Smoking history: Yes
- BMI: 33.8 (2023-06-01)

#### 3. Use Case 1: Encoder Vector

To generate a **patient embedding**, you can pass the entire block into an **LLM embedding model** (e.g., text-embedding-3-large, Gemma, or MedGemma 27B) using:

embedding = LLM\_embedder(text\_block)

This vector can then be:

- Clustered (e.g., for phenotyping)
- Used as input to a classifier
- Passed into a downstream TRD prediction model

# 4. Use Case 2: Risk Prediction Query Prompt

You can either use a **prompting format** for zero-shot/few-shot inference:

### ### Instruction:

You are a clinical assistant. Given the patient data below, estimate the risk of treatment-resistant depression.

#### ### Patient Information:

<Insert structured patient summary from above>

#### ### Answer:

Or build a **supervised training set** (serialized patient → TRD label) for fine-tuning a classifier.

# Notes on TRD Operationalization

- Often defined as **failure of ≥2 antidepressants** with adequate dose and duration
- More advanced models may incorporate:
  - Polypharmacy
  - Comorbid anxiety/PTSD
  - High PHQ-9 persistence
  - Lack of remission markers (e.g., continued suicidal ideation, functional impairment)



# Summary

#### Component **Purpose**

FHIR data → structured text Enables LLM-friendly serialization

Encoder vector Use embeddings for downstream ML/representation tasks

Estimate TRD risk from serialized text Prompt-based diagnosis

Optional fine-tuning Train classifier using embeddings + TRD outcome labels

The Art of War: In the midst of chaos, there is also opportunity. Martin

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