

Abstract screening report and full text download data preparation

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Summary

(All based on code below)

Tests

Valid screener notes?

Screener 1

- For screener 1, there is one “22”, which is marked as relevant. No correction needed.
 - Abstract: “We empirically investigate asset substitution of economic sectors in response to either liquidity demand or monetary policy shock, using the Japanese flow of funds data. Our interest is in responses of portfolio items of six sectors, financial institutions, central government, public corporations and local government, corporate business, personal, and overseas. It is reasonable to identify both shocks in liquidity demand and monetary policy, using recursive VAR models where the variable ordering is the real GDP, the GDP deflator, the commodity price index, the call

rate, the monetary base and each balance-sheet variable. We interpret that structural innovations in the monetary base as liquidity demand shocks and those in the call rate as monetary policy shocks. The responses to either the monetary-policy or liquidity demand shocks are different among sectors, among the transaction items, and between the periods before and after the bubble. Among the sectors, the personal sector exhibits the delayed responses, which are consistent with the limited participation model. The transaction items of the corporate business sector show vivid responses to the shocks. Between the two sub-periods, the stocks held by the financial institutions show clearly different responses.”

- For screener 1, there is one “00”, which is marked as irrelevant. No correction needed.
 - Abstract: “This article investigates the pricing behaviour of Turkish firms over the period 1988-2006 on the basis of firm-level micro data. The duration of prices is found to be 3.9 months on average. There is no clear heterogeneity across main groupings in the frequency of price changes, but more dependence on imported goods reduces price stickiness. Price decreases are less frequent than price increases, indicating downward rigidity in prices. There is evidence in favour of both time and state-dependent price setting behaviours. Further, there is a low degree of synchronization of price changes across firms, whereas price increases tend to be more synchronized than price decreases. Ordered probit models show that price adjustments depend on the type of the shock: the pass-through of a change in the cost is faster than changing demand. Besides, estimated probabilities of price adjustments with 5-years rolling windows reveal that inflation targeting has succeeded in bringing down the probability of price increases, whereas downward price rigidity has not weakened yet.”

Screener 2

- For screener 1, there is one “22”, which is marked as relevant. No correction needed.”
 - Abstract: “Existing theory and evidence on the effects of monetary policy are reviewed. Substantial room for disagreement among economists remains. New evidence, based on multivariate time series studies of several countries, is presented. While certain patterns in the data consistent with effective monetary policy are strikingly similar across countries, others, particularly the tendency of interest rate increases to predict high inflation, are harder to reconcile with effective monetary policy.”

Do screener notes correctly correspond to in-/exclusion of studies?

Screener 1

- For screener 1, there is one entry that was excluded though it carries an inclusion note. Correction not necessary because one mistake will only marginally affect the sample.
 - Abstract: “The Nigerian economy attracts abundance of foreign capital inflows and credit supply; hence, an adverse external credit shock might lead to a large decrease of external inflows due to global credit tightening, which may leave the domestic economy in deep recession. In this case, domestic monetary policy tools should be preferred to mitigate the external adverse effect on the domestic economy and stimulate investment. As a result, an important issue of concern in this study is how can the use of monetary policy tools mitigate the effect of external credit shocks on economic growth in Nigeria? In answering this question, this study attempted to assess the influence of monetary policy tools on external credit and economic growth nexus in Nigeria, using annual data covering 36 years for the period 1980–2015. The study adopted the Cobb–Douglas production function and estimated a specified model using autoregressive distributed lag cointegration approach. The study found out that cash reserve requirement, which is credit policy easing, is significant in growing the Nigerian economy, as compared to monetary policy rate. The implication of this is that, if credit policy easing is properly implemented, it could be efficient in offsetting adverse external credit shocks.”
- For screener 1, there is one entry that is included though it carries an exclusion note. But no correction necessary, as the entry will be excluded in the full text screening phase.
 - Abstract: “This paper analyzes the importance of monetary and fiscal policy shocks in explaining U.S. macroeconomic fluctuations, and establishes new stylized facts. The novelty of our empirical analysis is that we jointly consider both monetary and fiscal policy, whereas the existing literature only focuses on either one or the other. Our main findings are twofold: fiscal shocks are relatively more important in explaining medium cycle fluctuations whereas monetary policy shocks are relatively more important in explaining business cycle fluctuations, and failing to recognize that both monetary and fiscal policy simultaneously affect macroeconomic variables might incorrectly attribute the fluctuations to the wrong source.”

Screener 2

- For screener 2, there are 3 entries that were excluded though they carries inclusion notes. Correction not necessary because this mistake will only marginally affect the sample.
- Abstracts:
 1. “This study investigates the effect of monetary policy shocks in New Zealand and Australia on their respective exchange rates from 1985 to 1998 using vector autoregression

methodology. The results show that monetary policy shocks do contribute to the variability of both exchange rates, but these movements are not always consistent with theory. In particular, there is little support for the overshooting hypothesis. Also the results show that the exchange rates do not always move in the direction normally anticipated, particularly for New Zealand. A contraction in monetary policy may lead to a depreciation of the domestic currency rather than an appreciation.”

2. “This South African case study controls for the fiscal side of the economy using government borrowing as a potential accelerator of asymmetry in a monetary function that follows Taylor’s rule. Through the linear and non-linear ARDL framework, we find significant asymmetry effects of monetary policy on output and inflation, respectively. We also find government borrowing as an important underlying source of asymmetries in the response of macroeconomic fundamentals to monetary policy shocks in South Africa. Thus, we recommend that monetary authorities consider not only the effectiveness or otherwise of monetary policy instruments to affect the target policy goals, but also the fact that not all the target variables react in a similar way to expansionary and contractionary monetary policy shocks.”
3. “The perception that inflation dynamics in Sub-Saharan Africa (SSA) are driven by supply shocks implies a limited role for monetary policy in influencing inflation in the short run. SSA’s rapid growth, its integration with the global economy, changes in the policy frameworks, among others, in the last decade suggest that the drivers of inflation may have changed. We quantitatively analyze inflation dynamics in SSA using a Global VAR model, which incorporates trade and financial linkages among economies, as well as the role of regional and global demand and inflationary spillovers. We find that in the past 25 years, the main drivers of inflation have been domestic supply shocks and shocks to exchange rate and monetary variables; but that, in recent years, the contribution of these shocks to inflation has fallen. Domestic demand pressures as well as global shocks, and particularly shocks to output, however, have played a larger role in driving inflation over the last decade. We also show that country characteristics matter—the extent of oil and food imports, vulnerability to weather shocks, economic importance of agriculture, trade openness and policy regime, among others, help in explaining the role of shocks.”

- No invalid inclusions for screener 2.

Tests before and after merging screener data

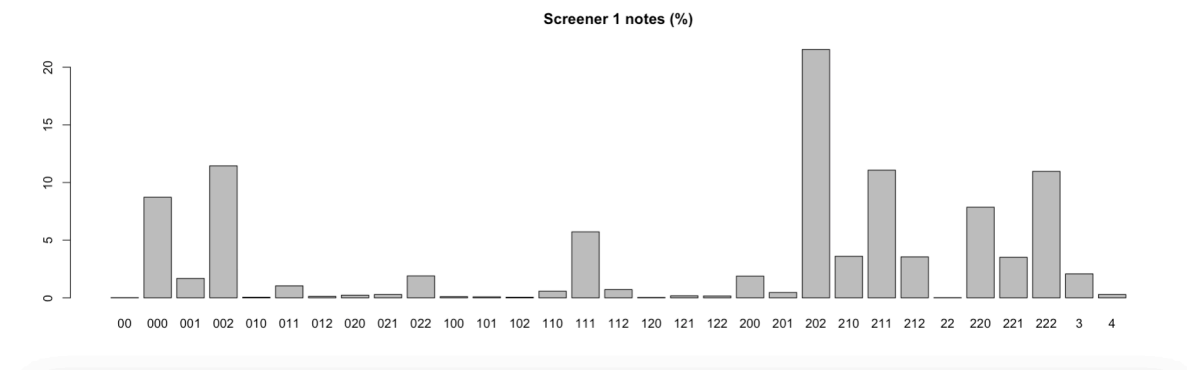
- “record_id” and “key” are correctly matching for both Screener 1 and Screener 2 data (i.e. ids and keys were not corrupted during data transformations and processing)
- merged data correctly matches original screener data (i.e. merging code is valid)

Analyses of screening

- 5826 entries were screened by either screener 1 or 2 (or both). 2034 were included by screener 1. 2148 were included by screener 2.
- 2540 were marked as relevant by at least one screener.

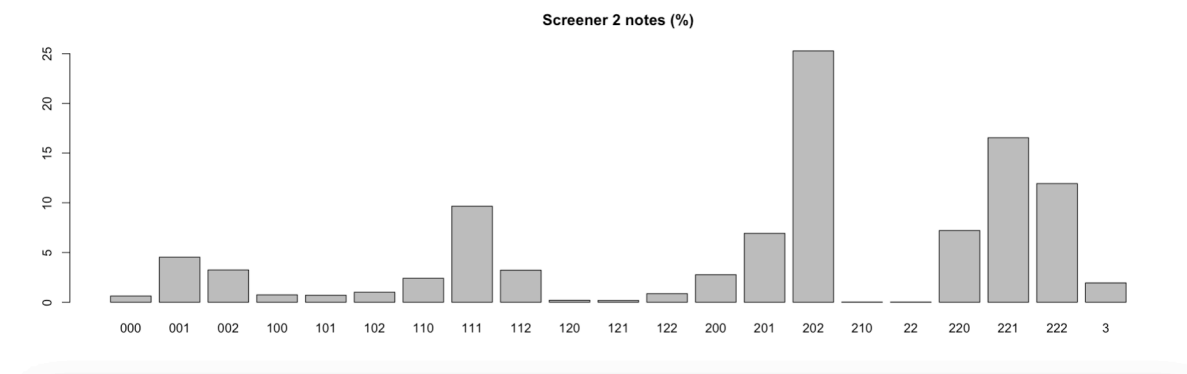
Screener 1 proportions of in-/exclusions notes

Screener 1 notes (%)



Screener 2 proportions of in-/exclusions notes

Screener 2 notes (%)



Agreement and overlap

- 1642 entries were included by both screeners.
- 2500 entries were excluded by both screeners.
- 335 entries were included by screener 1 but excluded by screener 2.

- 477 entries were included by screener 2 but excluded by screener 1.
- 4954 entries were screened by both screeners. Of these, 1977 , i.e. 39.9 %, were included by Screener 1 and 2119 , i.e. 42.8 %, were included by Screener 2.
- 726 entries were screened exclusively by Screener 1. Of these, 57 were included, i.e. 7.9 %.
- 146 entries were screened exclusively by Screener 2. Of these, 29 were included, i.e. 19.9 %.

Bibtex-Key creation

- There are 44 duplicate Bibtex-keys which need to be corrected when downloading the PDFs.
- There are also quite a few Bibtex-keys which have a bibtex-incompatible syntax. These would also need to be corrected.

Randomization and packages creation for full text download

- Order of entries has been randomized
- Data is split into a number `.xlsx` files with max of 100 entries in each file.
`.xlsx` files contain only information relevant for retrieving PDFs and empty columns for notes regarding availability.

Code and for testing, merging, analyzing and preparing full text download of abstract screening dataset

```
# library(readr)
library(tidyverse)
# library(openxlsx)
library(here)

# Importing screener 1 data (all)
s1_data <- read_csv(here("data/study search/database search/processed/post_AS/screener_1_d

# Importing screener 2 data (all)
s2_data <- read_csv(here("data/study search/database search/processed/post_AS/screener_2_d

## Some consistency tests before merging
# Screener 1: only valid screener notes?
unique(s1_data$exported_notes_1)
```

```

paste('For screener 1, there is one "22", which is marked as relevant. No correction needed.')
View(s1_data %>% filter(exported_notes_1=="22")) # Inspecting the entry.
paste(s1_data %>% filter(exported_notes_1=="22") %>% select(`abstract note`)) # Inspecting
paste('For screener 1, there is one "00", which is marked as irrelevant. No correction needed.')
View(s1_data %>% filter(exported_notes_1=="00")) # Inspecting the entry.
paste(s1_data %>% filter(exported_notes_1=="00") %>% select(`abstract note`)) # Inspecting
paste("Exclusion seems valid.")
# Screener 2: only valid screener notes?
unique(s2_data$exported_notes_1)
paste('For screener 1, there is one "22", which is marked as relevant. No correction needed.')
View(s2_data %>% filter(exported_notes_1=="22")) # Inspecting the entry.
paste(s2_data %>% filter(exported_notes_1=="22") %>% select(`abstract note`)) # Inspecting

# Do screener 1 notes correctly correspond to in-/exclusion of studies?
inclusion_notes <- c("111", "112", "121", "211", "122", "212", "221", "222") # Index of entries
s1_included_notes <- s1_data$exported_notes_1 %in% inclusion_notes
sum(s1_data$included[s1_included_notes] != 1)
paste("For screener 1, there is one entry that was excluded though it carries an inclusion note.")
View(s1_data %>% filter(included == "0" & exported_notes_1 %in% inclusion_notes)) # Inspecting
paste(s1_data %>% filter(included == "0" & exported_notes_1 %in% inclusion_notes) %>% select(`abstract note`))

exclusion_notes <- c("000", "001", "002", "010", "011", "012", "020", "021", "022", "100", "101", "102", "110", "111", "112", "120", "121", "122", "200", "201", "202", "210", "211", "212", "220", "221", "222")
s1_excluded_notes <- s1_data$exported_notes_1 %in% exclusion_notes
sum(s1_data$included[s1_excluded_notes] != 0)
paste("For screener 1, there is one entry that is included though it carries an exclusion note.")
View(s1_data %>% filter(included == 1 & exported_notes_1 %in% exclusion_notes)) # Inspecting
paste(s1_data %>% filter(included == 1 & exported_notes_1 %in% exclusion_notes) %>% select(`abstract note`))

# Do screener 2 notes correctly correspond to in-/exclusion of studies?
s2_included_notes <- s2_data$exported_notes_1 %in% inclusion_notes
sum(s2_data$included[s2_included_notes] != 1)
paste("For screener 2, there are 3 entries that were excluded though they carry an inclusion note.")
View(s2_data %>% filter(included == "0" & exported_notes_1 %in% inclusion_notes)) # Inspecting
paste(s2_data %>% filter(included == "0" & exported_notes_1 %in% inclusion_notes) %>% select(`abstract note`))

s2_excluded_notes <- s2_data$exported_notes_1 %in% exclusion_notes
sum(s2_data$included[s2_excluded_notes] != 0)
paste("No invalid inclusions for screener 2.")

## Merging the data
# Test that record id and key are matching for both s1_data and s2_data

```

```

all(sort(paste0(s1_data$record_id, s1_data$key)) == sort(paste0(s2_data$record_id, s2_data

# Merge s1_data and s2_data based on all variables that must have equal values, i.e. those
columns <- colnames(s1_data) # These are the column names of the dataframe (they are equal
matching_columns <- head(columns, -3) # These are the columns that have not been created b
merged_data <- merge(s1_data, s2_data, by = matching_columns, all = T, sort = F, suffixes
# Test that merged data correctly matches original data
all(na.omit(merged_data[, c("record_id", "key", "included.s1", "exported_notes_1.s1")]) ==
from_merged <- na.omit(merged_data[, c("record_id", "key", "included.s2", "exported_notes_
from_merged <- from_merged[order(from_merged$record_id),]
from_s2 <- na.omit(s2_data[, c("record_id", "key", "included", "exported_notes_1")])
from_s2 <- from_s2[order(from_s2$record_id),]
all(from_merged == from_s2) # For screener 2, test that record_id, key, included and export

# Analyses of screening
s1_s2_agreement_data <- merged_data %>%
  select(key, included.s1, included.s2, exported_notes_1.s1, exported_notes_1.s2) %>%
  filter(!is.na(included.s1) | !is.na(included.s2))
paste(nrow(s1_s2_agreement_data), "entries were screened by either screener 1 or 2.", sum(
paste(nrow(s1_s2_agreement_data %>% filter(included.s1==1 | included.s2==1)), "were marked
paste(nrow(s1_s2_agreement_data %>% filter(included.s1==1 & included.s2==1)), "were marked
# Screener 1 proportions of in-/exclusions notes
round(prop.table(table(s1_s2_agreement_data$exported_notes_1.s1)) * 100, 1)
barplot(prop.table(table(s1_s2_agreement_data$exported_notes_1.s1)) * 100)
title("Screener 1 notes (%)")
# Screener 2 proportions of in-/exclusions notes
round(prop.table(table(s1_s2_agreement_data$exported_notes_1.s2)) * 100, 1)
barplot(prop.table(table(s1_s2_agreement_data$exported_notes_1.s2)) * 100)
title("Screener 2 notes (%)")
# Agreement and overlap
# Included by both screeners
included_both <- s1_s2_agreement_data %>%
  filter(included.s1 == 1 & included.s2 == 1)
paste(nrow(included_both), "entries were included by both screeners.")
# Excluded by both screeners
excluded_both <- s1_s2_agreement_data %>%
  filter(included.s1 == 0 & included.s2 == 0)
paste(nrow(excluded_both), "entries were excluded by both screeners.")
# Included by screener 1, excluded by screener 2
included_s1_excluded_s2 <- s1_s2_agreement_data %>%
  filter(included.s1 == 1 & included.s2 == 0)

```



```

paste(nrow(included_s1_excluded_s2), "entries were included by screener 1 but excluded by
# Included by Screener 2, excluded by screener 1
included_s2_excluded_s1 <- s1_s2_agreement_data %>%
  filter(included.s1 == 0 & included.s2 == 1)
paste(nrow(included_s2_excluded_s1), "entries were included by screener 2 but excluded by
# Screened by both screener 1 and 2
both <- s1_s2_agreement_data %>%
  filter(!is.na(included.s2) & !is.na(included.s1))
paste(nrow(both), "entries were screened by both screeners. Of these,",
  sum(both$included.s1), ", i.e.", round(sum(both$included.s1)/nrow(both)*100,1), "%",
# Exclusively screened by Screener 1
only_s1 <- s1_s2_agreement_data %>%
  filter(is.na(included.s2) & !is.na(included.s1))
paste(nrow(only_s1), "entries were screened exclusively by Screener 1. Of these,",
  sum(only_s1$included.s1), "were included, i.e.", round(sum(only_s1$included.s1)/nrow
# Exclusively screened by Screener 2
only_s2 <- s1_s2_agreement_data %>%
  filter(is.na(included.s1) & !is.na(included.s2))
paste(nrow(only_s2), "entries were screened exclusively by Screener 2. Of these,",
  sum(only_s2$included.s2), "were included, i.e.", round(sum(only_s2$included.s2)/nrow

# Extracting relevant studies
relevant_entries <- merged_data %>%
  filter(included.s1 == 1 | included.s2 == 1)
# Check that number of included in merged relevant entries matches number of included in o
sum(relevant_entries$included.s1, na.rm = T) == sum(s1_data$included, na.rm = T)
sum(relevant_entries$included.s2, na.rm = T) == sum(s2_data$included, na.rm = T)

# Adding a bibtex key

#' Creates a Bibtex key from a dataframe with publication information.
#'
#' This function takes three inputs, year, author and
#' title, and creates a Bibtex key for each row in the dataframe. The Bibtex key has
#' the format: name-of-first-author_year_first-word-of-title.
#'
#' @param year A vector with the publication years.
#' @param author A vector with the authors' names.
#' @param title A vector with the publication titles.
#' @return A vector with the Bibtex keys.
create_bibtex_key <- function(year, author, title) {

```

```

# Split author names and select first name
first_author <- strsplit(author, ",")[[1]][1]
# Remove non-alphanumeric characters from first word of title and convert to lowercase
first_word <- tolower(gsub("[^:alnum:]", "", strsplit(title, " ")[[1]][1]))
# Combine components to create Bibtex key
paste0(first_author, "_", year, "_", first_word)
}
relevant_entries$BibtexKey <- apply(relevant_entries, 1, function(row) {
  create_bibtex_key(row["publication year"], row["author"], row["title"])
})
# Test uniqueness of bibtex keys:
if (length(unique(relevant_entries$BibtexKey)) == length(relevant_entries$BibtexKey)) {
  message("The generated Bibtex-keys are unique.")
} else {
  warning("There are ", length(relevant_entries$BibtexKey) - length(unique(relevant_entries$BibtexKey)),
    " duplicated keys. View(as.data.frame(relevant_entries$BibtexKey[duplicated(relevant_entries$BibtexKey)]))")
}

# Randomizing the order of the dataset
# Set the random seed
set.seed(364)
# Randomizing the dataframe
randomized_relevant_entries <- relevant_entries[sample(nrow(relevant_entries)),]

# Extract data for full text download
data_for_download <- randomized_relevant_entries %>%
  select(record_id, key, author, title, `publication year`, `publication title`, issue, volume)
  mutate(most_recent_version_available = NA, not_available = NA, retracted = NA, duplicate = NA)

# Create packages for full text download
#' Save Dataframe to Multiple Excel Files
#'
#' This function saves the entries of a dataframe into xlsx files with a maximum of 100 entries per file.
#'
#' @param df The dataframe to be saved.
#' @param file_path The path and filename prefix for the output files.
#'
#' @return None
#'
#' @export
save_to_xlsx <- function(df, file_path) {

```

```

# Get number of files to be saved
n_files <- ceiling(nrow(df) / 100)

# Split dataframe into chunks of 100 rows
df_list <- split(df, rep(1:n_files, each=100, length.out=nrow(df)))

# Save each chunk to a separate file
for (i in seq_along(df_list)) {
  file_name <- paste0(file_path, "_", i, ".xlsx")
  write.xlsx(df_list[[i]], file_name, rowNames = FALSE)
}

}
# save_to_excel(data_for_download, here("data/study search/database search/processed/post_A

```