

Modelling with Impact:

What you need to know as an Analytics Professional

Craig Savage, PhD, MBA

Citizens Data Scientists - Melbourne 27 Feb, 2018





 Share highlights of my experiences in accelerating algorithm development projects

"Intelligence is learning from your mistakes, wisdom is learning from the mistakes of others." "Wise men learn from others' mistakes, fools from their own."





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- Case study
 - Banking provisions under a new accounting standard (IFRS9)
 - Variety of tools required to solve effectively



Introduction

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- Case study
 - Banking provisions under a new accounting standard (IFRS9)
 - Variety of tools required to solve effectively
- Highlights for "What you need to know as an Analytics Professional" throughout the talk.

Example problem: Implement IFRS9 Accounting Standard

- Introductions to....
 - Credit Risk
 - IFRS 9
- Modelling procedure
 - Data
 - Procedures
 - Review

Example problem: Implement IFRS9 Accounting Standard

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- Introductions to....
 - Credit Risk
 - <u> IEDC O</u>

Talk will be at a high level: Majority of the talk will be Introduction!

Review





- Loans are profitable when customers repay, lose money otherwise
 - <u>Provisions</u> are used to cover expected losses
 - Expected Credit Loss: $P_D \times LGD \times EAD$
 - Collective Provisions are allotted against all loans
 - Specific Provisions are raised against doubtful/impaired loans
 - In theory, Specific Provisions = Collective Provisions = Expected Credit Loss
 - Provisions count as an *Expense* (i.e. lowers profits)
- <u>Advanced Banks</u> maintain their own models for P_D , LGD, EAD
- A portfolio of loans is a book (e.g. mortgage book)
- A loan that increases in risk and recovers is said to cure.



Single-Slide Introduction: Credit Risk

- Loans are profitable when customers repay, lose money otherwise
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Highly idealised case: Single-Slide Introduction!

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What you need to know as an Analytics Professional

- Business jargon is like a language
 - Easier to understand/read than speak/write
 - Similar to swearing to use effectively, need a certain proficiency
- How to learn:
 - Internet (Blogs, Wikipedia)
 - LinkedIn for mentors in the area
- Three main reasons you'd be brought on:
 - Lack of capacity: They know what needs to be done, understand the complexity, don't have time.
 - Lack of capability: They have the time, but don't have the knowledge.

What you need to know as an Analytics Professional

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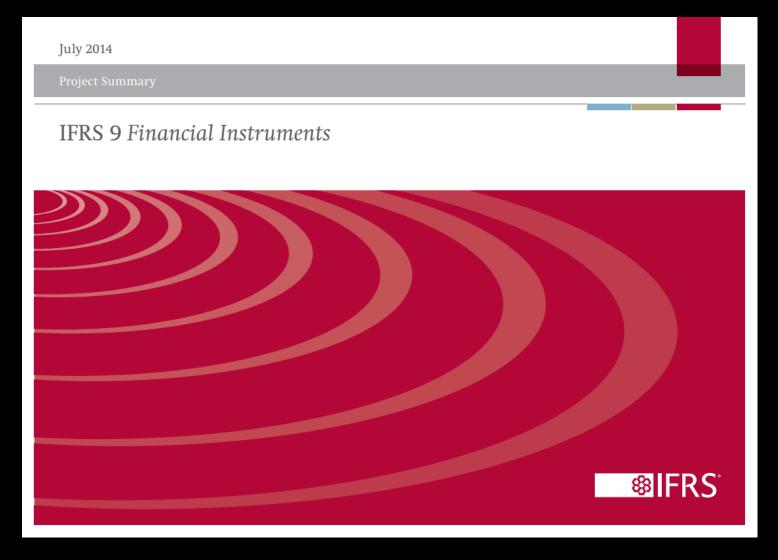
- Business jargon is like a language
 - Easier to understand/read than speak/write

The Dreaded Both: They have neither time nor knowledge

"Using Agile, we'll implement Machine Learning with Big Data on Azure to improve our NPS on Social Media with Millennials!"

- Lack of capacity: They know what needs to be done, understand the complexity, don't have time.
- Lack of capability: They have the time, but don't have the knowledge.

- Summary of rules online
 - Powerpoint as Documentation



July 2

Impairment

IFR

A forward-looking impairment model



Why is the IASB addressing impairment?

During the financial crisis, the delayed recognition of credit losses on loans (and other financial instruments) was identified as a weakness in existing accounting standards. Specifically, the existing model in IAS 39 (an 'incurred loss' model) delays the recognition of credit losses until there is evidence of a trigger event. This was designed to limit an entity's ability to create hidden reserves that can be used to flatter earnings during bad times.

As the financial crisis unfolded, it became clear that the incurred loss model gave room to a different kind of earnings management, namely to postpone losses. Even though IAS 39 did not require waiting for actual default before impairment is recognised, in practice this was often the case.

The complexity of IAS 39, which used multiple impairment models for financial instruments, was also identified as a concern.

How will the new requirements improve financial reporting?

The main objective of the new impairment requirements is to provide users of financial statements with more useful information about an entity's expected credit losses on financial instruments. The model requires an entity to recognise expected credit losses at all times and to update the amount of expected credit losses recognised at each reporting date to reflect changes in the credit risk of financial instruments.

This model is forward-looking and it eliminates the threshold for the recognition of expected credit losses, so that it is no longer necessary for a trigger event to have occurred before credit losses are recognised. Consequently, more timely information is required to be provided about expected credit losses.

Furthermore, when credit losses are measured in accordance with IAS 39, an entity may only consider those losses that arise from past events and current conditions. The effects of possible future credit loss events cannot be considered, even when they are expected. The requirements in IFRS 9 broaden the information that an entity is required to consider when determining its expectations of credit losses.

Specifically, IFRS 9 requires an entity to base its measurement of expected credit losses on reasonable and supportable information that is available without undue cost or effort, and that includes historical, current and forecast information.

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July 2

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Overview of the impairment requirements



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As soon as a financial instrument is originated or purchased, 12-month expected credit losses are recognised in profit or loss and a loss allowance is established.

Stage 1

This serves as a proxy for the initial expectations of credit losses.

For financial assets, interest revenue is calculated on the gross carrying amount (ie without adjustment for expected credit losses).

What are the stages?

Stage 2

If the credit risk increases significantly and the resulting credit quality is not considered to be low credit risk, full lifetime expected credit losses are recognised.

Lifetime expected credit losses are only recognised if the credit risk increases significantly from when the entity originates or purchases the financial instrument.

The calculation of interest revenue on financial assets remains the same as for

Stage 3

If the credit risk of a financial asset increases to the point that it is considered credit-impaired, interest revenue is calculated based on the amortised cost (ie the gross carrying amount adjusted for the loss allowance). Financial assets in this stage will generally be individually assessed.

Lifetime expected credit losses are still recognised on these financial assets.

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16 | IFRS 9 Financial Instruments | July 2014

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Measuring expected credit losses



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What should an entity consider when measuring expected credit losses?

Credit losses are the present value of all cash shortfalls. Expected credit losses are an estimate of credit losses over the life of the financial instrument. When measuring expected credit losses, an entity should consider:

- (a) the probability-weighted outcome: expected credit losses should represent neither a best or worst-case scenario. Rather, the estimate should reflect the possibility that a credit loss occurs and the possibility that no credit loss occurs:
- (b) the time value of money: expected credit losses should be discounted to the reporting date;
- (c) reasonable and supportable information that is available without undue cost or effort.

What information is used?

An entity is required to use reasonable and supportable information that is available at the reporting date without undue cost or effort, and that includes information about past events, current conditions and forecasts of future conditions.

IFRS 9 does not prescribe particular measurement methods. Also, an entity may use various sources of data that may be internal (entity-specific) and external.

Entities are not required to use a 'crystal ball' to predict the future; what an entity uses depends on the availability of information. As the forecast horizon increases, it is expected that the specificity of information used to measure expected credit losses will decrease. (For example, rather than estimating specific cash flow shortfalls it may be necessary to consider information such as historical loss rates adjusted as relevant for current and forecast conditions).

Although the model is forward-looking, historical information is always considered to be an important anchor or base from which to measure expected credit losses. However, historical data should be adjusted on the basis of current observable data to reflect the effects of current conditions and forecasts of future conditions.

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What you need to know as an Analytics Professional



- Learn about your business area
- Make a checklist of requirements for your solution:
 - Data: "...reasonable and supportable information that is available without undue cost or effort"
 - Time value of money need a discount rate.
 - Need to classify Stage 1, Stage 2, Stage 3.
 - Need to forecast economic conditions
 - Migration between stages
 - No requirements on algorithms, types, etc.
 - Must be fun to audit!
 - Ask 10 analysts, get 10 different solutions

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"... reasonable and supportable information that is available without undue cost or effort"

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 Banking key metrics available from the Australian Prudential Regulatory Authority (APRA)

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• Banking ke Statistics Regulatory

Quarterly ADI Performance

September 2017 (released 5 December 2017)

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Contents Important notice

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 Banking ke Regulatory

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Oua Revisions

Septeml This edition of the publication may contain revisions to previously published statistics. Significant revisions, if any, are identified and quantified in the 'Important notice'.

This publication includes revisions to previously published statistics if better source data are available or if compilation errors are uncovered.

APRA regularly analyses past revisions to identify potential improvements to the source data and statistical compilation techniques, in order to minimise the frequency and scale of any future revisions.

Forthcoming issues

This publication will be released according to the timetable published on the APRA website.

Notation

Amounts are expressed in millions of Australian dollars. Both the Australian-dollar denominated transactions and the Australian-dollar equivalent of foreign-currency denominated transactions are included.

The symbol "" indicates that the data have been masked to maintain confidentiality.

The blank cells represent items that are not applicable, for example where data were collected up to or from a certain period end date.

Glossary and explanatory notes

A set of explanatory notes is provided at the end of the publication to assist the reader in understanding the source and definitions of the data. In particular, these notes help explain differences between the data presented and information publicly released by banks in their financial statements and profit

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Thanks to APRA for making the data freely available. I acknowledge their copyright to the data.

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Table 5d Major banks' asset quality (\$ million, consolidated Sep 2004 Jun 2005 Sep 2005 Sep 2008 Dec 2008 Impaired facilities Non-accrual items with provisions 2,231 2,133 2,011 1,749 19,476 20,162 677 708 739 604 729 772 612 1.019 1,264 1,483 Non-accrual items without provisions 152 18 17 Restructured items with provisions 175 171 124 73 72 102 Restructured items without provisions 1,514 Other real estate owned 11 30 Other assets acquired through security enforcement Total Impaired facilities 3,435 3,011 2,905 2,885 2,634 2,597 2,607 2,420 2,340 2,599 2,820 6,829 11,846 14,985 18,377 19,241 20,291 22,365 23,180 of which: Facilities in Australia 2,040 1,861 2,173 5,060 16,227 16,310 Provisions held 1,345 1,341 1,094 4,339 5,753 6,925 6,381 7,330 7,483 Specific provisions 1,555 1,495 1,378 4,387 8,936 Security held 1,309 5,876 7,116 9,901 11,964 13,115 Impaired facilities to loans and advances: Total provisions held 2,900 2,837 2,347 2,403 3,54 11,628 14,041 15,317 16,642 19,294 20,599 of which: Facilities in Australia 1,745 1,754 1,743 2,810 9,311 10,766 11,583 12,301 14,571 14,677 Stage 3 Past due items 1,963 2,263 3,448 3,409 3,893 8,010 8,733 8,503 9,203 9,490 10,516 of which: Facilities in Australia 1,659 1,976 2,789 2,744 3.042 3,57 6.734 6,727 7,944 8,859 5,398 Sum of impaired facilities and past due items 9,311 27,745 31,855 Specific Provisions to impaired facilities: Gross loans and advances 811,907 1,278,885 1,659,038 LGD (about 40%) Impaired facilities to loans and advances 0.4% 0.4% 0.2% 0.4% 1.2% 0.2% Pillar Banks 0.3% 0.3% 0.5% Past due to loans and advances 0.2% 0.3% 0.3% 0.3% 0.6% 0.6% 42.0% 33.2% Specific provisions to impaired facilities 39.1% 41.9% 42.1% 36.6% 34.3% 37.5% 40.1% 33.2% 32.8% 32.3% Specific provisions and security held to impaired facilities 88.7% 91.7% 88.5% 85.6% 82.3% 80.4% 81.6% 74.29 88.9% Number of entities

Past due to loans and advances: Stage 2

ory notes

provided at the end of the publication to assist the reader in understanding the data. In particular, these notes help explain differences between the on publicly released by banks in their financial statements and profit

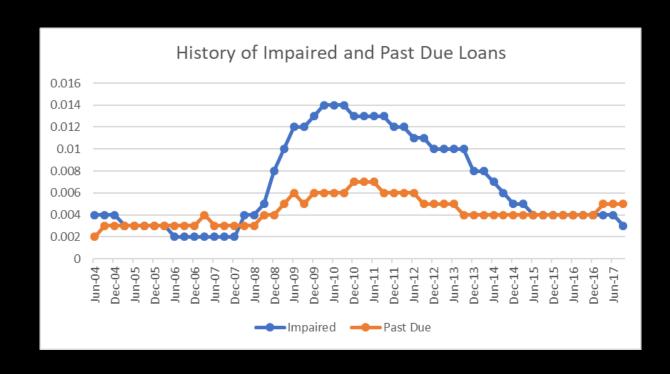
Data Visualisation: Target Variables

- Visual representation of data
 - Trends, periods, outliers
- How you present matters
 - Graph types
 - Graph theme
- Visualisations demonstrate your Tribe

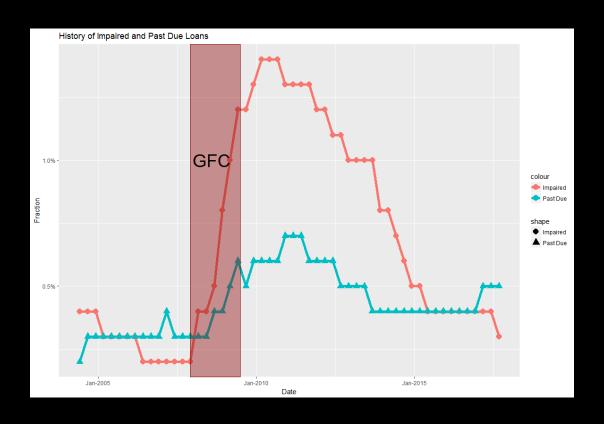
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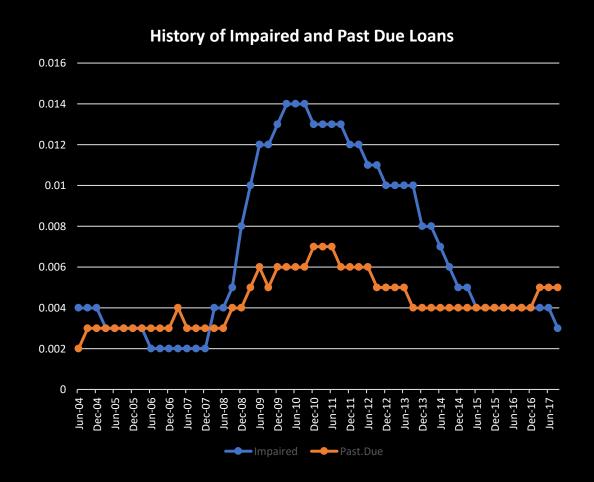


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Data Visualisation: Target Variables

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- Australian Bureau of Statistics has a "Modeller's Database" of 119 variables
 - That's more variables than we have historical data from APRA



ime Series Workbook

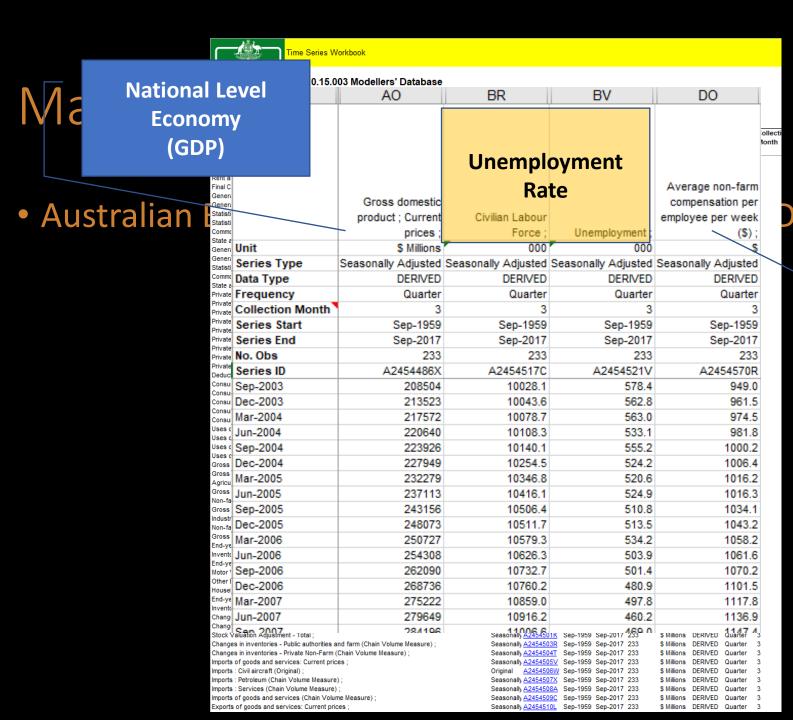
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Related Information: <u>Summary Publication</u>	Inquiries								
Data Item Description	Series Type	Series ID	Series Start	Series End	No. Obs.	Unit	Data Type	Freq.	Collecti Month
Rent and other dwelling services: Current prices ;	Seasonally	A2454446F	Sep-1959	Sep-2017	233	\$ Millions	DERIVED	Quarter	3
Final Consumption Expenditure: Current prices :		A2454447J		Sep-2017	233	\$ Millions	DERIVED	Quarter	3
Rent and other dwelling services: Chain volume measures ;		A2454448K		Sep-2017			DERIVED	Quarter	3
Final Consumption Expenditure: Chain volume measures ;		A2454449L		Sep-2017		\$ Millions	DERIVED	Quarter	3
General government - National; Final consumption expenditure: Current prices;		A2454450W		Sep-2017		\$ Millions	DERIVED	Quarter	3
General government - State and local; Final consumption expenditure; Current prices;		A2454451X		Sep-2017	233	S Millions	DERIVED	Quarter	3
Statistical Discrepancy (E);		A2454452A	Sep-1959		233	S Millions	DERIVED	Quarter	3
Statistical Discrepancy (I);		A2454453C		Sep-2017		S Millions	DERIVED	Quarter	3
Commonwealth Government Gross Fixed Capital Formation ;	Seasonally	A2454454F	Sep-1959	Sep-2017	233	\$ Millions	DERIVED	Quarter	3
State and Local Government Gross Fixed Capital Formation ;		A2454455J		Sep-2017		\$ Millions	DERIVED	Quarter	3
General government - National; Final consumption expenditure (Chain Volume Measure);	Seasonally	A2454456K	Sep-1959	Sep-2017	233	\$ Millions	DERIVED	Quarter	3
General government - State and local; Final consumption expenditure (Chain Volume Measure)	Seasonally	A2454457L	Sep-1959	Sep-2017	233	\$ Millions	DERIVED	Quarter	3
Statistical Discrepancy (E) (Chain Volume Measure);	Seasonally	A2454458R	Sep-1959	Sep-2017	233	\$ Millions	DERIVED	Quarter	3
Commonwealth Government Gross Fixed Capital Formation (Chain Volume Measure);	Seasonally	A2454459T	Sep-1959	Sep-2017	233	\$ Millions	DERIVED	Quarter	3
State and Local Government Gross Fixed Capital Formation (Chain Volume Measure);	Seasonally	A2454460A	Sep-1959	Sep-2017	233	\$ Millions	DERIVED	Quarter	3
Private; Gross fixed capital formation - Dwellings - Total: Current prices;	Seasonally	A2454461C	Sep-1959	Sep-2017	233	\$ Millions	DERIVED	Quarter	3
Private; Gross fixed capital formation - Intellectual property products: Current prices;	Seasonally	A2454462F	Sep-1959	Sep-2017	233	\$ Millions	DERIVED	Quarter	3
Private; Gross fixed capital formation - Cultivated biological resources: Current prices;	Seasonally	A2454463J	Sep-1959	Sep-2017	233	\$ Millions	DERIVED	Quarter	3
Private; Gross fixed capital formation - Machinery and equipment - Total: Current prices;	Seasonally	A2454464K	Sep-1959	Sep-2017	233	\$ Millions	DERIVED	Quarter	3
Private; Gross fixed capital formation - Dwellings - Total (Chain Volume Measure);	Seasonally	A2454465L	Sep-1959	Sep-2017	233	\$ Millions	DERIVED	Quarter	3
Private; Gross fixed capital formation - Intellectual property products (Chain Volume Measure);	Seasonally	A2454466R	Sep-1959	Sep-2017	233	\$ Millions	DERIVED	Quarter	3
Private; Gross fixed capital formation - Cultivated biological resources (Chain Volume Measure)	Seasonally	A2454467T	Sep-1959	Sep-2017	233	\$ Millions	DERIVED	Quarter	3
Private; Gross fixed capital formation - Non-dwelling construction - Total (Chain Volume Measur	Seasonally	A2454468V	Sep-1959	Sep-2017	233	\$ Millions	DERIVED	Quarter	3
Private; Gross fixed capital formation - Machinery and equipment - Total (Chain Volume Measure	Seasonally	A2454469W	Sep-1959	Sep-2017	233	\$ Millions	DERIVED	Quarter	3
Deductions from Gross Rent;	Seasonally	A2454470F	Sep-1959	Sep-2017	233	\$ Millions	DERIVED	Quarter	3
Consumption of fixed capital - Non-Financial Corporations and Financial Corporations;	Original	A2454471J	Sep-1959	Sep-2017	233	\$ Millions	DERIVED	Quarter	3
Consumption of fixed capital - Dwellings Owned by Persons;	Original	A2454472K	Sep-1959	Sep-2017	233	\$ Millions	DERIVED	Quarter	3
Consumption of fixed capital - Private Financial Enterprises;	Original	A2454473L	Sep-1959	Sep-2017	233	\$ Millions	DERIVED	Quarter	3
Consumption of fixed capital - Farm Unincorporated Trading Enterprises;	Original	A2454474R	Sep-1959	Sep-2017	233	\$ Millions	DERIVED	Quarter	3
Consumption of fixed capital - Non-farm Unincorporated Trading Enterprises;	Original	A2454475T	Sep-1959	Sep-2017	233	\$ Millions	DERIVED	Quarter	3
Uses of income - Property income payable - Interest on dwellings ;	Seasonally	A2454476V	Sep-1959	Sep-2017	233	\$ Millions	DERIVED	Quarter	3
Uses of income - Property income payable - Total interest ;		A2454477W		Sep-2017		\$ Millions	DERIVED	Quarter	3
Uses of income - Property income payable - Property income payable by unincorporated enterpri				Sep-2017	233	\$ Millions	DERIVED	Quarter	3
Uses of income - Property income payable - Consumer debt interest;		A2454480K	Sep-1959			\$ Millions	DERIVED	Quarter	3
Gross agricultural value added at basic prices ;		A2454481L		Sep-2017		\$ Millions	DERIVED	Quarter	3
Gross agricultural value added at producers' prices ;		A2454482R		Sep-2017		\$ Millions	DERIVED	Quarter	3
Agriculture - Taxes less subsidies on products ;		A2454483T		Sep-2017	233	\$ Millions	DERIVED	Quarter	3
Gross Non-Farm Product at Factor Cost ;		A2454484V		Sep-2017		\$ Millions	DERIVED	Quarter	3
Non-farm; Gross domestic product; Current prices;		A2454485W		Sep-2017			DERIVED	Quarter	3
Gross domestic product ; Current prices ;		A2454486X		Sep-2017		\$ Millions	DERIVED	Quarter	3
Industry Gross Value Added (Chain Volume Measure) - Agriculture, forestry and fishing - Agricu				Sep-2017		\$ Millions	DERIVED	Quarter	3
Non-farm ; Gross domestic product (Chain Volume Measure) ;		A2454488C		Sep-2017		\$ Millions	DERIVED	Quarter	3
Gross domestic product (Chain Volume Measure);		A2454489F		Sep-2017		\$ Millions	DERIVED	Quarter	3
End-year net capital stock Non-Financial Corporations and Financial Corporations; Current prices				Sep-2017		\$ Millions	DERIVED	Quarter	3
Inventories - Private Non-Farm ; Current prices ;		A2454492V		Sep-2017		\$ Millions	DERIVED	Quarter	3
End-year net capital stock Non-Financial Corporations and Financial Corporations (Chain Volume		A2454493W		Sep-2017	233	\$ Millions	DERIVED	Quarter	3
Motor Vehicles (Chain Volume Measure);		A2454494X		Sep-2017	233	\$ Millions	DERIVED	Quarter	3
Other Building and Structures: Total (Chain Volume Measure);		A2454495A	Sep-1959			\$ Millions		Quarter	3
Household Durables (Chain Volume Measure);		A2454496C	Sep-1959			\$ Millions	DERIVED	Quarter	3
End-year net capital stock - Machinery and Equipment : Total (Chain Volume Measure) ;		A2454497F		Sep-2017		\$ Millions	DERIVED	Quarter	3
Inventories - Private Non-Farm (Chain Volume Measure);		A2454498J		Sep-2017		\$ Millions	DERIVED	Quarter	3
Changes in inventories - Public authorities and farm;		A2454499K		Sep-2017		\$ Millions	DERIVED	Quarter	•
Changes in inventories - Private Non-Farm ; Current prices ;		A2454500J A2454501K		Sep-2017		\$ Millions	DERIVED DERIVED	Quarter	3
Stock Valuation Adjustment - Total ;		A2454501K A2454503R		Sep-2017		\$ Millions		or or or to	3
Changes in inventories - Public authorities and farm (Chain Volume Measure);				Sep-2017	233	\$ Millions	DERIVED	Quarter	3
Changes in inventories - Private Non-Farm (Chain Volume Measure);		A2454504T		Sep-2017	233	\$ Millions	DERIVED	Quarter	•
Imports of goods and services: Current prices ;		A2454505V	Sep-1959			\$ Millions		Quarter	3
Imports : Civil aircraft (Original) ;	Original		Sep-1959			\$ Millions	DERIVED	Quarter	3
Imports : Petroleum (Chain Volume Measure) ;		A2454507X	Sep-1959			\$ Millions	DERIVED	Quarter	3
Imports : Services (Chain Volume Measure) ;		A2454508A		Sep-2017	233	\$ Millions	DERIVED	Quarter	3
Imports of goods and services (Chain Volume Measure);		A2454509C	Sep-1959				DERIVED	Quarter	3
Exports of goods and services: Current prices ;	Seasonally	A2454510L	Sep-1959	Sep-2017	233	\$ Millions	DERIVED	Quarter	3

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Wage Growth

Thanks to the ABS for making the data freely available. I acknowledge their copyright to the data.

I have no affiliation with the ABS.

Although I'm using their data, that should not be construed as their endorsement of my work.

Stock Valuation Adjustment - Total Seasonally A2454501K Sep-1959 Sep-2017 233 \$ Millions DERIVED Quarter Changes in inventories - Public authorities and farm (Chain Volume Measure): Seasonally A2454503R Sep-1959 Sep-2017 233 \$ Millions DERIVED Quarter Changes in inventories - Private Non-Farm (Chain Volume Measure); Sep-1959 Sep-2017 233 mports of goods and services: Current prices; Sep-1959 Sep-2017 233 Original A2454506W Sep-1959 Sep-2017 233 mports : Civil aircraft (Original) Imports : Petroleum (Chain Volume Measure) Sep-1959 Sep-2017 233 \$ Millions DERIVED Quarter Imports : Services (Chain Volume Measure): Imports of goods and services (Chain Volume Measure) Seasonally A2454509C Sep-1959 Sep-2017 233 \$ Millions DERIVED Quarter Exports of goods and services: Current prices Seasonally A24545101 Sen-1959 Sen-2017 233 S Millions DERIVED Quarter

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What you need to know as an Analytics Professional

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- Learn about your business area
- Make a checklist of requirements for your solution
- Data is never perfect
 - Where did it come from?
 - Is it documented? Or is there someone you can ask about it?
 - Where is your threshold for "good enough data"?
 - Do you have what you need to achieve project goals?
 - Does the client understand the limitations of the data and what your algorithm cannot reasonably be expected to do?

Are they limited in capacity or capability?

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Modelling procedure

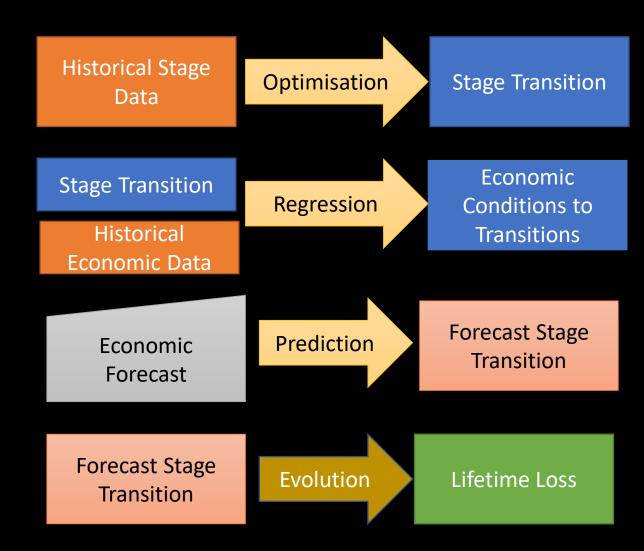
Remember – high level!

If you're an analyst and you get annoyed, imagine how the accountants would feel when I went through all of IFRS 9 in a few slides!

If you're an accountant feeling annoyed, imagine how the Credit Risk Analysts would feel when I went through Credit Risk in 1 slide!

Modelling Procedure: End-to-End Plan

- Need to forecast economic conditions
 - Arbitrary horizon
 - Requires human input can't rely solely on historical data!
- Migration between stages
 - Contingent upon forecast economic conditions



Modelling Procedure: Stage Transition

- Have historical data of fraction of loans in each Stage
- Estimate a transition matrix based on observed distribution
- For simplicity:
 - Everything entering Stage 3 produces a loss equal to the LGD
 - Migration between Stages i and j given by $p_{i\,i}$
- For Evolution $s_{t+1} = Ts_t$

$$\underline{T} = \begin{bmatrix} p_{11} & p_{21} \\ p_{12} & p_{22} \\ p_{13} & p_{23} \end{bmatrix}$$

$$\hat{s}_{t+1} = Ts_t$$
 For Stages 1 & 2: $\arg\min_{\underline{T}} \|s_{t+1} - \hat{s}_{t+1}\|_2$ Subject to: $0 < p_{ij} < 1$
$$p_{11} + p_{12} + p_{13} = 1$$

$$p_{21} + p_{22} + p_{23} = 1$$

Modelling Procedure: Regression

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Regression theory:

$$y = f(x; \beta) + \epsilon$$

Find suitable f.

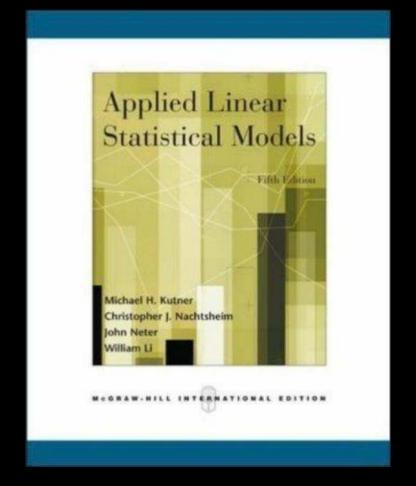
- Methods:
 - Pick a functional form, optimize parameters

$$y = \underline{X}\beta + \epsilon$$
$$y = \frac{\beta_0 \exp(\beta_1 x)}{1 + \exp(\beta_1 x)} + \epsilon$$

SVM, Neural Networks, LOESS, ...

I used linear regression

$$p_{ij}(t) \approx \beta_0 + \beta_1 u(t) + \beta_2 \Delta G(t) + \beta_3 \Delta w(t)$$



Modelling Procedure: Economic Forecasts

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- ARIMA models: Auto-Regressive Integrated Moving Average
- Vector Auto-Regressive (VAR)
- Diagnostics for "best" model fits
 - Variety of cost functions: Different cost functions imply different models are "best"!
- Can't use historical data alone, due to IFRS 9 requirements



Search

Home » Forecasting: principles and practice » 8 ARIMA models

8 ARIMA models

ARIMA models provide another approach to time series forecasting. Exponential smoothing and ARIMA models are the two most widely-used approaches to time series forecasting, and provide complementary approaches to the problem. While exponential smoothing models were based on a description of trend and seasonality in the data, ARIMA models aim to describe the autocorrelations in the data.

Before we introduce ARIMA models, we need to first discuss the concept of stationarity and the technique of differencing time series.

- 8.1 Stationarity and differencing
- 8.2 Backshift notation
- 8.3 Autoregressive models
- 8.4 Moving average models
- 8.5 Non-seasonal ARIMA models
- 8.6 Estimation and order selection
- 8.7 ARIMA modelling in R
- 8.8 Forecasting
- 8.9 Seasonal ARIMA models
- 8.10 ARIMA vs ETS
- 8.11 Exercises
- 8.12 Further reading

Book information

About



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Rob J Hyndman George Athanasopoulos

Forecasting: principles and practice

- Getting started
- The forecaster's toolbox
- Judgmental forecasts
- Simple regression
- ▶ Multiple regression
- Time series decomposition
- ▶ Exponential smoothing
- → ARIMA models
- Stationarity and differencing
- Backshift notation

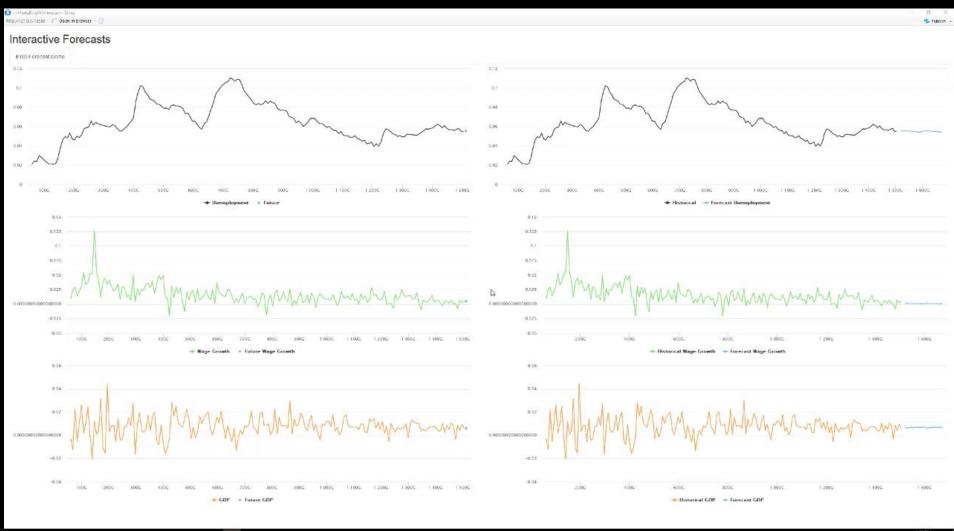
< 7.9 Further reading

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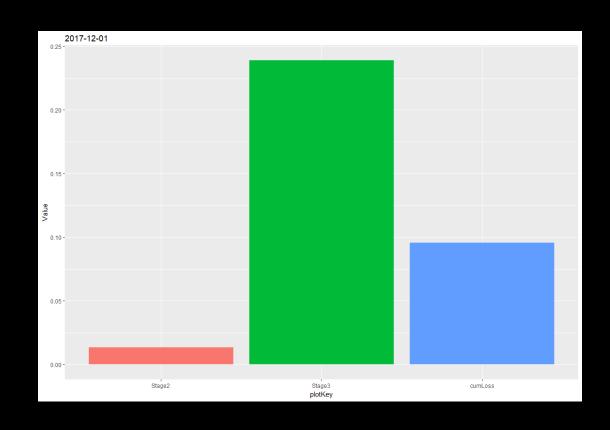
8.1 Stationarity and differencing

• Buckstille flotatio

Modelling Procedure: Economic Forecasts



- Only concerned with the part of the portfolio in Stage 2
 - Impulse response: Start with the whole in Stage 2, see how it evolves.
 - Transition matrices calculated based on regression and forecast information
- From Stage 3, 40% (i.e. LGD) is added to cumulative losses
- Results in estimate for lifetime loss.



Modelling Procedure: Evolution

- Only concerned with the part of the portfolio in Stage 2
 - Impulse response: Start with the whole in Stage 2, see how it evolves.
 - Transition matrices calculated based on regression and forecast information
- From Stage 3, 40% (i.e. LGD) is added to cumulative losses
- Results in estimate for lifetime loss.



What you need to know as an Analytics Professional

- Learn about your business area
- Make a checklist of requirements for your solution
- Data is never perfect
- Variety of tools/methods
 - Choice: Specialist or Generalist?
 - Choice: Focus on Algorithms or Problems?
 - Algorithms vs Problem
 - Kalman filters, Wavelets, SVM, k-means, ARIMA, Fuzzy Logic, Graph Theory
 - Tracking, Feature Extraction, Classification, Clustering, Forecasting, Fusion, Assignment
- Stakeholder engagement/ownership

- Everything on the requirements done
 - Cumulative loss by month need discount factor, but easily calculated
 - Forecast based on Expert Input + VAR models
 - Transition based on forecast macro-economics
- Review: Present to stakeholders



• Host of technical assumptions, questions, problems: What training/test data methodology did you use?

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• Host of technical assumptions, questions, problems: What training/test data methodology did you use?

What tests for appropriateness did you use for your VAR?

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Host of technical assumptions, questions, problems:
 What training/test data methodology did you use?

What tests for appropriateness did you use for your VAR?

What other regression options did you investigate?

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Host of technical assumptions, questions, problems:
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Host of technical assumptions, questions, problems:
 What training/test data methodology did you use?

What's the Impact?

What other regression options did you investigate?

Impacts: Case-Study Ernst & Young Survey

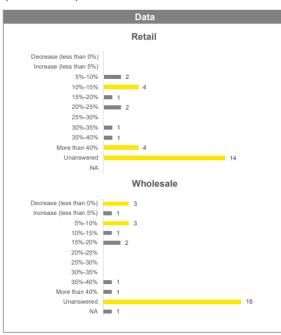


Impacts: Case-Study Ernst & Young Survey





1. Impact assessment – impairment provisions Expected percentage increase in total impairment provisions on transition to IFRS 9 (continued)



Commentary

Credit card exposures driving increase in retail provisions

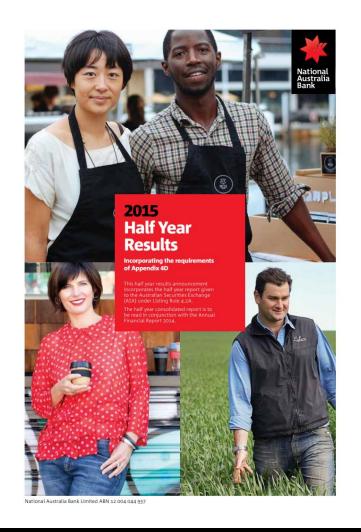
- Retail portfolios will be most impacted by the adoption of IFRS 9, in particular, because of exposures classified as stage 2, and resulting lifetime expected credit loss (ECL) requirements.
- ► The highest impact has been reported on credit card portfolios, with four banks expecting more than 40% increase in provisions. This is largely due to the requirement to calculate ECL for both the drawn and undrawn exposures.
- Some banks reported a significant impact on unsecured products, especially with the introduction of a 12-month expected loss for stage 1 exposures.
- ▶ There is more diversity on the impact of mortgages: four banks expect an increase of 5%-10%; three an increase of more than 40%; one a decrease.

Wholesale impact expected to be less significant

- ► The expected impact on wholesale portfolios is generally lower than the impact on retail portfolios, with the exception of "central governments and central banks" and "financial institutions", where it appears more significant as they currently attract no, or only small, provisions. However, the absolute impact is expected to be small due to the high quality of these assets.
- ▶ The introduction of 12-month ECL for financial instruments that are considered to have low credit risk contributes to the increase in wholesale provisions.
- Some banks noted little change, or even a decrease, in ECL for corporates, primarily resulting from relatively long emergence period used under IAS 39, but also because of very high credit quality or significant collateral. This is more evident for countries where larger collective provisions were being booked on watch list exposures under IAS 39.
- Some corporate assets were also reclassified to fair value through profit and loss, which are not subject to credit impairment.

"There is more diversity on the impact of mortgages: four banks expect an increase of 5%-10%; three an increase of more than 40%; one a decrease."

EY IFRS 9 Impairment Banking Survey





Review of Operating Environment, Group Operations and Results

2015

This was driven by:

- Lower specific charges for business lending and the unsecured retail portfolio in Australian Banking
- Recoveries in UK Banking reflecting an improvement in the UK economy
- Further reductions to NAB UK CRE impaired assets.

The March 2015 half year collective provision B&DD charge was \$49 million. This includes an overlay for Australian agriculture and resource sectors as market conditions continued to tighten in these segments.

Provisions for Bad and Doubtful Debts

Total provisions for B&DDs increased by \$261 million to \$4,351 million during the March 2015 half year, mainly due to an increase in collective provisions of \$783 million on transition to AASB 9. This was partially offset by a decrease in NAB UK CRE provisions (£262 million) resulting from loan sales in the half year.

Specific provisions decreased by \$547 million to \$907 million for the March 2015 half year, largely driven by the disposal of the NAB UK CRE loans (£177 million) and continued improvement in the performance of the Australian Banking impaired portfolio.

The Group coverage of specific provisions to gross impaired assets increased marginally from 35.3% at September 2014 to 35.5% at March 2015.

Total collective provisions have increased since September 2014 to \$3.44 million at March 2015. Excluding the impact of AASB 9 transitional adjustments and foreign exchange, there was an underlying reduction of \$48 million. This was mainly due to the sale of NAB UK CRE loans which resulted in a collective provision release, partly offset by an overlay for the agriculture and resource sectors in Australia and an increase in the potential credit exposure on the Group's derivatives portfolio driven by mark-to-market movements.

The collective provision to credit risk weighted assets ratio has increased by 18 basis points from 0.83% at September 2014 to 1.01% at March 2015. Excluding the impact of AASB 9 transitional adjustments, the collective provision coverage was seven basis points lower largely due to the sale of NAB IUK CRE loans in the half.

90+ Days Past Due and Gross Impaired Assets

	As at		
	Mar 15	Sep 14	Mar 14
90+ days past due loans (\$m)	2,292	2,342	2,526
Gross impaired assets (\$m)	2,558	4,122	5,614
90+ days past due and gross			
impaired assets (\$m)	4,850	6,464	8,140

	As at		
	Mar 15	Sep 14	Mar 14
90+ days past due loans to gross loans and acceptances	0.40%	0.43%	0.47%
Gross impaired assets to gross loans and acceptances	0.45%	0.76%	1.05%
90+ days past due and gross impaired assets to gross loans and			
acceptances	0.85%	1.19%	1.52%

Non-Impaired Assets 90+ Days Past Due

The Group ratio of 90+ days past due loans to gross loans and acceptances (90+ DPD ratio) declined by three basis points to 0.40% during the March 2015 half year, mainly due to the sale of NAB UK CRE loans in the half. In addition, improvements were observed in Group's business lending and retail lending 90+ DPD ratios.

Gross Impaired Assets

The Group ratio of gross impaired assets to gross loans and acceptances (impaired asset ratio) decreased 31 basis points to 0.45% during the March 2015 half year. The improvement was mainly driven by the business lending portfolio, reflecting the sale of NAB UK CRE impaired assets (£430 million) and a general decrease in impairment activity for Australian Banking and UK Banking.

Net Write-Offs

The Group's net write-offs to gross loans and acceptances annualised ratio decreased by nine basis points to 0.22% during the March 2015 half year. The decline was experienced across all major operating regions, excluding UK Banking.

The 12 month rolling net write-off rate for the Group's retail portfolio decreased two basis points to 0.12% over the six months to March 2015 and declined by one basis point to 0.03% for the Group's housing portfolio.

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National Australia Bank Limited ABN 12 004 044 937

Review of Operating Environment, Group Operations and Results

Half Year Results 2015

This was driven by:

- Lower specific charges for business lending and the unsecured retail portfolio in Australian Banking
- Recoveries in UK Banking reflecting an improvement in the UK economy
- Further reductions to NAB UK CRE impaired assets.

The March 2015 half year collective provision B&DD charge was \$49 million. This includes an overlay for Australian agriculture and resource sectors as market conditions continued to tighten in these segments.

Provisions for Bad and Doubtful Debts

Total provisions for B&DDs increased by \$261 million to \$4,951 million during the March 2015 half year, mainly due to an increase in collective provisions of \$783 million on transition to AASB 9. This was partially offset by a decrease in NAB UK CRE provisions (£262 million) resulting from loan sales in the half year.

Specific provisions decreased by \$547 million to \$907

million for the March 2015 ha the disposal of the NAB UK (and continued improvement Australian Banking impaired

The Group coverage of spec impaired assets increased m September 2014 to 35.5% at

Total collective provisions hav September 2014 to \$3,444 mi Excluding the impact of AASE and foreign exchange, there of \$48 million. This was main UK CRE loans which resulted release, partly offset by an ov and resource sectors in Austra the potential credit exposure controlled this payment. In mark to me controlled the payment to me.

The collective provision to cre ratio has increased by 18 bas September 2014 to 1.01% at impact of AASB 9 transitional provision coverage was sever due to the sale of NAB UK CF

90+ Days Past Due and Assets

90+ days past due loans (\$m) Gross impaired assets (\$m)

90+ days past due and gross impaired assets (\$m)

Non-Impaired Assets 90+ Days Past Due

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Net Write-Offs

"Total provisions for [Bad and Doubtful Debts] increased by \$261 million to \$4,351 million during the March 2015 half year, mainly due to an increase in collective provisions of \$783 million on transition to AASB 9."

past due loans to gross			
and acceptances	0.40%	0.43%	0.47
paired assets to gross loans ceptances	0.45%	0.76%	1.05
past due and gross ed assets to gross loans and			
ances	0.85%	1.19%	1.529





Review of Operating Environment, Group Operations and Results

This was driven by

- Lower specific charges for business lending and the
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The March 2015 half year collective provision B&DD charge was \$49 million. This includes an overlay for Australian agriculture and resource sectors as market conditions continued to tighten in these segments

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Total collective provisions h September 2014 to \$3,444 m Excluding the impact of AA of \$48 million. This was mail UK CRE loans which resul release, partly offset by an ov and resource sectors in Aust the potential credit exposur

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90+ Days Past Due and

90+ days past due loans (\$m)

90+ days past due and gros

Non-Impaired Assets 90+ Days Past Due

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Gross Impaired Assets

The Group ratio of gross impaired assets to gross loans and acceptances (impaired asset ratio) decreased by 31 basis points to 0.45% during the March 2015 half year lending portfolio, reflecting the sale of NAB UK CRE impaired assets (£430 million) and a general decrease

Net Write-Offs

"Total provisions for [Bad and Doubtful Debts] increased by \$261 million to \$4,351 million during the March 2015 half year, mainly due to an increase in collective provisions of \$783 million on transition to AASB 9."

NAB IFRS 9 **Impact**

(As at June, 2015, that was about US\$590million)

90+ days past due loans to gros Gross impaired assets to gross loans 0.45% 0.76% 90+ days past due and gross impaired assets to gross loans and 0.85% 1.19% 1.52%

National Australia Bank Limited ABN 12 004 044 937





National Australia Bank Limited ABN 12 004 044 937

Review of Operating Environment, Group Operations and Results

- Lower specific charges for business lending and the
- Recoveries in UK Banking reflecting an improvement Further reductions to NAB UK CRF impaired assets.
- The March 2015 half year collective provision B&DD

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The collective provision to cr ratio has increased by 18 bas September 2014 to 1.01% a impact of AASB 9 transitions provision coverage was seve due to the sale of NAB UK C

90+ Days Past Due and

90+ days past due loans (\$m)

90+ days past due and gross impaired assets (\$m)

90+ days past due loans to gross loans and acceptances Gross impaired assets to gross loans 0.45% 0.76% 90+ days past due and gross impaired assets to gross loans and 0.85% 1.19% 1.52%

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Non-Impaired Assets 90+ Days Past Due

Gross Impaired Assets

The Group ratio of gross impaired assets to gross loans and acceptances (impaired asset ratio) decreased by 31 basis points to 0.45% during the March 2015 half year lending portfolio, reflecting the sale of NAB UK CRE impaired assets (£430 million) and a general decrease

Net Write-Offs

"Total provisions for [Bad and Doubtful Debts] increased by \$261 million to \$4,351 million during the March 2015 half year, mainly due to an increase in collective provisions of \$783 million on transition to AASB 9."

https://en.wikipedia.org/wiki/List of countries by GDP (nominal) Per the United Nations, 2016

192	Grenada	884
193	Saint Kitts and Nevis	852
194	The Gambia	851
195	Samoa	824
196	Vanuatu Vanuatu	812
197	Turks and Caicos Islands	797
198	Saint Vincent and the Grenadines	729
199	Comoros	648
200	■ Dominica	533
201	Tonga	435
202	São Tomé and Príncipe	337
203	Federated States of Micronesia	308

(As at June, 2015, that was about US\$590million)

NAB IFRS 9 **Impact**



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- *Modelling with Impact* doesn't refer to:
 - Jargon
 - Presentation Style
 - Colour schemes
 - Interactive plots
 - Animated GIFs
- If your model goes live tomorrow, what do you expect to happen?

$$\sum_{t=1}^{20} FV(t) \exp(-r_d t)$$

With a 1%/quarter discount rate (~4.1% annual)



Citizen
Data
Scientists
– Melbourne

- *Modelling with Impact* doesn't refer to:
 - Jargon
 - Presentation Style
 - Colour schemes
 - Interactive plots
 - Animated GIFs
- If your model goes live tomorrow, what do you expect to happen?

$$\sum_{t=1}^{20} FV(t) \exp(-r_d t)$$

With a 1%/quarter discount rate (~4.1% annual)

A\$2.7Billion

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- Citizens Data Science Meetup: Microsoft & Sudarshan Roy
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- Awkward Taylor Swift Dancing Tumblr

Where you can find me

YouTube

https://goo.gl/Mjp6Sc

GitHub

https://github.com/SavageDoc/