

Competing for your member's business during uncertain times with AI

How well is your credit union positioned?



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- **Telephone:** +1 646 558 8656

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Slide Link

Today's slides can be found online at:
<http://bit.ly/2022-07-20-zest>

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Use the

Questions Box

located on the bottom of your screen to
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Please take our post-event survey. We
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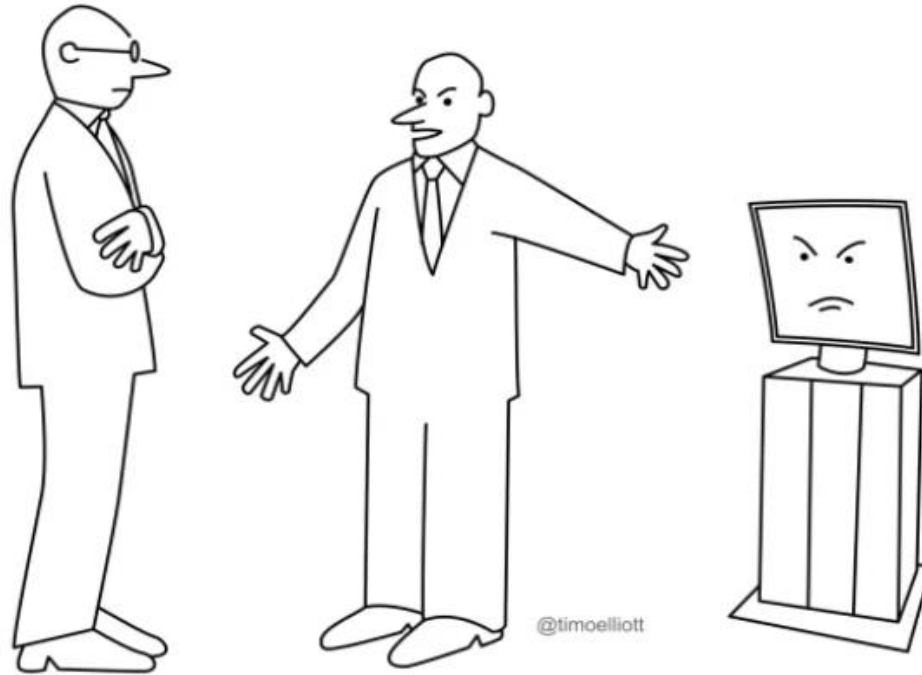
Agenda

- Demystifying AI/ML
- About Zest AI
- Credit Union Objectives and Challenges
- Process in Building Auto Decisioning Model
- ABC Credit Union Model Result Example
- Q&A

Demystifying AI and ML

Automated Decisioning



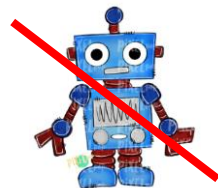


*His decisions aren't any better than yours
— but they're WAY faster...*



AI decisioning is not....

- Is not a physical robot for the processing of loans
- Is not a technology that allows for continuous learning with a disregard to compliance policies
- Is not technology that is intended to replace human interaction with members



What is AI and ML?



AI (Artificial Intelligence)

The use of faster computing to process massive data and compute more algorithms (better math), to get better insights on data and then act on it.

Machine Learning

Using computing to apply algorithms that are able to learn things from past use cases without programming.

Automation (AI) examples

- Travel booking system 1985 (DOT)
- Airport Check-in Kiosk 1995
- IBM Deep Blue vs Garry Kasparov 1997 (FIDE)
- Phone book replaced by search engines
- Apple Watch with fall detection



Big computers processing tons of DATA with better math and compliant

About Zest AI



Who is Zest AI?

We are a CUSO and our mission is to make fair and transparent credit available to all of your members.

Our Automated underwriting platform enables Credit Unions to deploy and monitor powerful, compliant underwriting models swiftly and easily to benefit your members.

History of Zest

- Founded in 2009 by Google CIO
- Hired data scientist to build AI Models for lending
- Pivoted in 2016 focusing on banks, implementing customized AI Models with tools for the data scientist
- Worked closely with the regulatory agencies



DISCOVER



citi



FIFTH THIRD BANK

TRUIST 

Credit Union Pivot

- In 2019 Mike de Vere was announced CEO
- Making fair and transparent credit available to the underserved became Zest's core mission
- Strategy was to leverage the technology developed for banks and make Auto Decisioning (AI) available, scalable, and **affordable** for Credit Unions



Zest Credit Union Pivot



- May 2021 Zest began delivering tailored AI Underwriting Models based on specific membership data and portfolios
- Consulting fees were waived for the Proof-of-Concept (POC) model engagement
- Zest revised the pricing model for Credit Unions

Automated Decisioning Early Adopters

- 2021 40+ Credit Unions
- 2021 CURQL Investment
- 2022 Additional 80-100 Clients
- 2022 CUSO of the Year



Zest's New Clients

MAX Credit Union

Copper State Credit Union

Proceed Finance

Beehive Federal Credit Union

Achieva Credit Union

Redstone Federal Credit Union

GreenState Credit Union

SkyOne Federal Credit Union

Commonwealth Credit Union

First Service Credit Union

MIDFLORIDA Credit Union

PrimeWay Federal Credit Union

Alabama Credit Union

Mainstreet Credit Union

Service Credit Union

Community Choice Credit Union

Golden 1 Credit Union

Verity Credit Union

Clark County Credit Union

TruWest Credit Union

Kinecta Federal Credit Union

Space Coast

FirstLight Federal Credit Union

Farmers Insurance Federal
Credit Union

Pearl Hawaii FCU

Maps Credit Union

Evansville Teachers Federal
Credit Union

Credit Union of Texas

All In Credit Union

Altra Federal Credit Union

Delivering 100's Compliant AI Models

- Zest was the first company to **testify before the Congressional Artificial Intelligence Task Force**
- The Federal Reserve **cites our work**, refers to Zest as “gold standard” and hosted Zest on the Fed’s podcast
- The OCC adopted Zest talking points in its 2019 report on **risks in the banking system** (see pg 23)
- Zest has trained top bank examiners at FDIC, OCC, Fed Reserve, CFPB, and FHFA on how to **examine ML models**
- The National Fair Housing Alliance **endorsed our ML underwriting tools** to the nation’s top bank CEOs

Regulators deem Zest the

“Gold Standard in Machine Learning”



cfpb

NFHA
National Fair Housing Alliance



Best in adding value to members, creates an innovative solution to a problem, reduces operational costs and increases income for all credit unions.

C[★] CUSO
of the year!

- Collaboration
- Innovation
- Cooperative heart
- Business acumen
- Inspirational
- Future-focused

Credit Union Objectives and Challenges



Credit Union Core Objectives

Expand Credit & Drive Loan Growth

Increase lending volume and take rates



Safely Manage & Reduce Risk

Reduce loan losses and enable risk-based pricing



Increase Automation & Efficiency

Increase automated decisioning, streamline compliance



Enhance Market Competitiveness

Innovate quickly, identify and respond to trends faster

Drive Fairer Outcomes

Improve fair lending outcomes and lend more inclusively

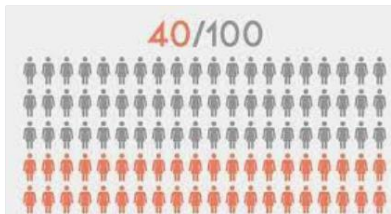


Improve Borrower Experience

Increase member satisfaction



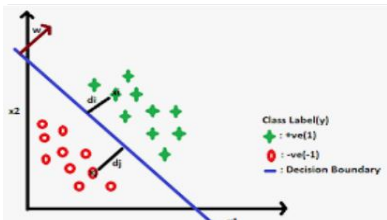
Today's Credit Union Challenges



40%

Of Americans are difficult for lenders to assess accurately

Source: CFPB and Experian



20-40

Year old scoring techniques are out of date



83%

Credit unions take longer than 30 minutes to approve or deny a member loan

Source: Zest NACUSO Survey

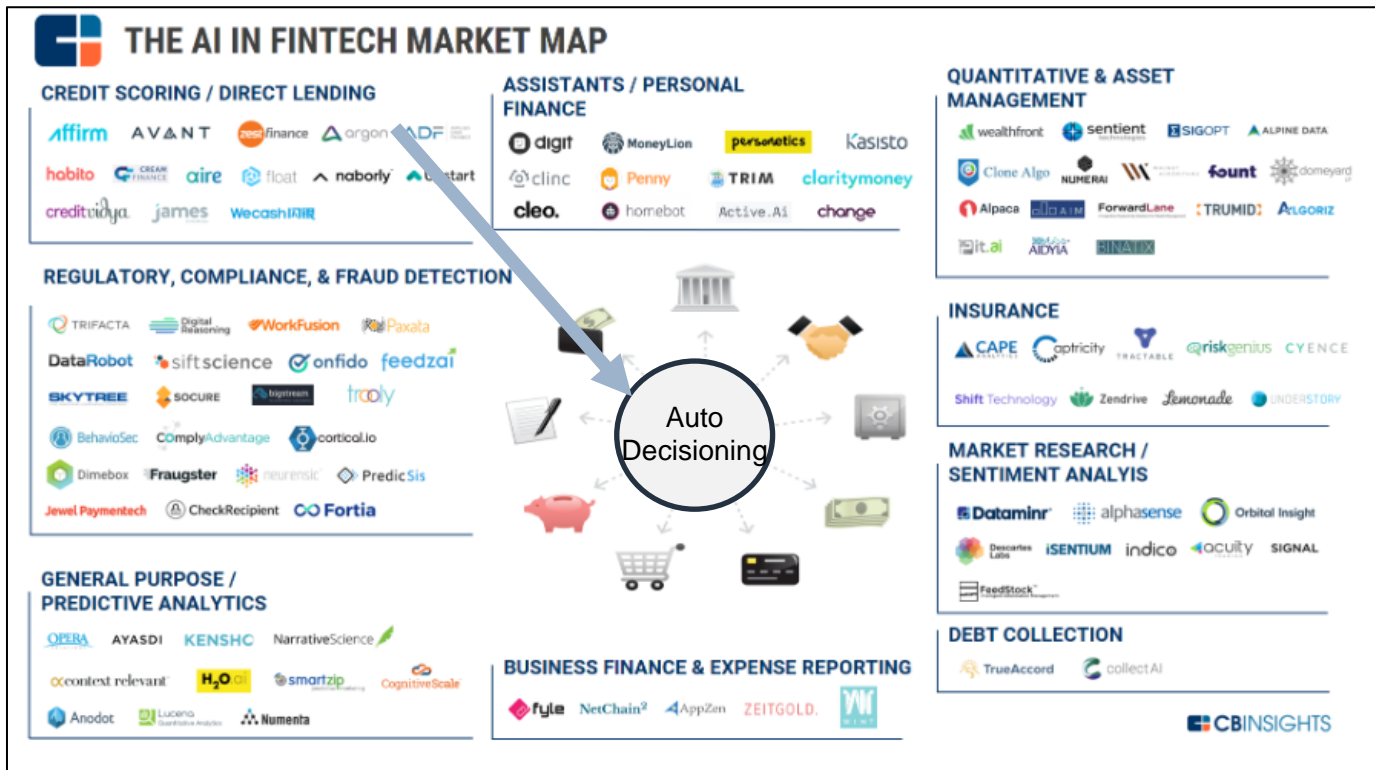


50%

Fintech share of personal loans business

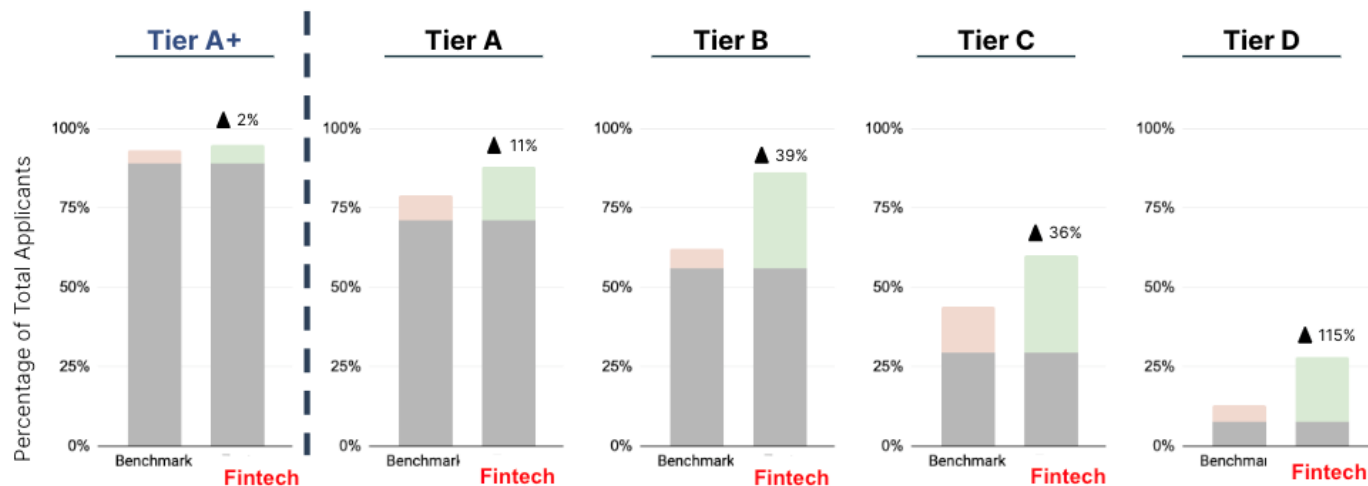
Source: Experian and Allied Market Research

How are Fintechs doing it?



What tiers are Fintechs targeting?

SWAPSET RESULTS BY CREDIT TIER



20-30%

→ Overall approval rate increase versus Benchmark

LEGEND In-In Swap-Out Swap-In

Process in Building Auto Decisioning Models to Compete



Data required to build a model

Organizational Questions										
1. Credit Bureau used:										
2. States the Client is operates in (or would like to operate in the future):										
3. Business goal (Reduce losses, increase approvals, automation)										
4. Current model / scorecard										
a. In house Custom? If yes, Logistic RRegression? App features/non-std Bureau data?										
b. Vendor Custom? If yes, Logistic Regression? Who? App features/non-std Bureau data?										
c. Primarily FICO Score										
d. Primarily Vantage Score										
e. Other: _____										
5. Current LOS:										
6. Product Specific Credit score distribution (if custom feel free to create own matrix):										
Product Line Information Direct Auto		Portfolio Level Statistics	Credit Tier Stratification (Use whatever makes sense)							
			Credit Tier A+	Credit Tier A	Credit Tier B	Credit Tier C	Credit Tier D	Credit Tier E	Credit Tier F	NO SCORE
Portfolio Total			Superprime	Prime	Prime	Near Prime	Subprime			
FICO/Vantage Range										
Total Applications (#)										
Auto Approvals (#)										
Auto Denials (#)										
Total Approved (#)										
Total Funded (#)										
Total Origination (\$)										
Average Funded Amount										
# of loans charged off										
Average Amount Charged off (\$) per Bad										
Average Value of a Good Loan										
Average Interest rate charged										
Average Loan Term										
Average Loan life										

Modeling starts with a LOT of data



Pool of data from the relevant bureau

- National Sample
- Multiple Loan Products
- All Credit Tiers

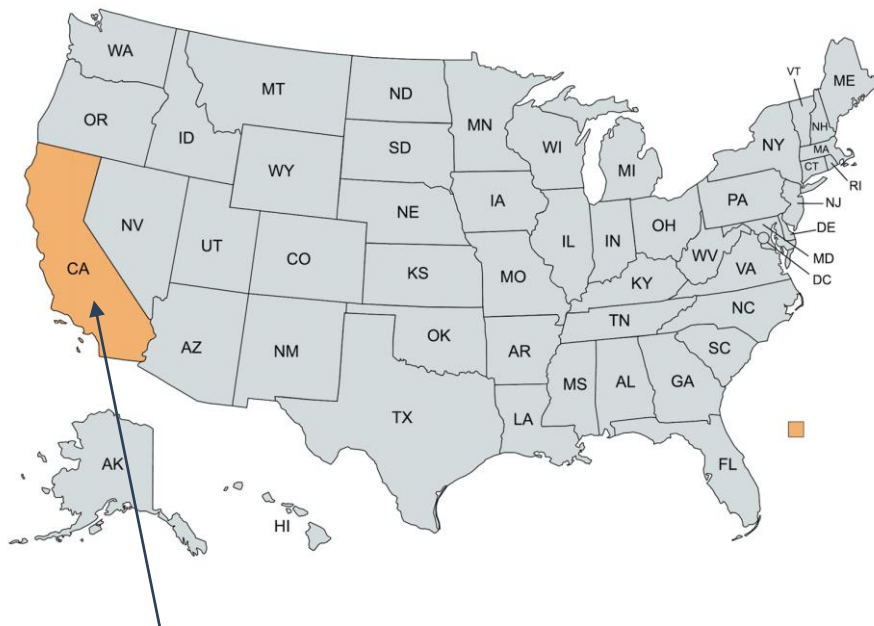


TransUnion^{tu}

experian.

EQUIFAX

Preliminary Model



Profile of Current + Future Customers

Pool narrowed down to your:

- Geography
- Specific loan product(s)
- Credit tiers distribution

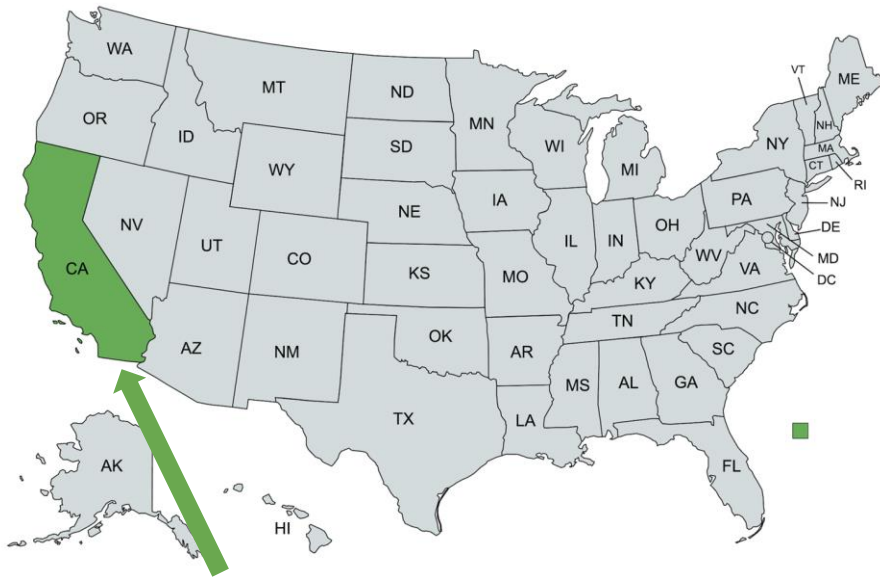
Built using a large, representative sample of lookalike bureau data

The lookalike sample will mirror the demographics of the communities served today and hope to serve in the future

Using a lookalike sample allows the model to learn from millions of records

Preliminary model performance will be compared to FICO / Vantage Scores

Model optimized & validated with CU specific data



Member historic data included for:

- Validation
- Optimization



Lender specific data

The final model will include bureau sourced loan performance data (including reject inference) & member application data



Fully compliant MRM including

economic benefit, swap set analysis, fair lending analysis, and deployment plan



Full access to Model Management System

The final model is available within the MMS in 30-45 days following receipt of lenders data.

Status Quo Score vs. Zest Score

ML consistently outperforms traditional methods

Traditional Scorecards /Logistical Regression	Machine Learning / Artificial Intelligence
Limited to 10-50 variables One-way static	Use 100s or 1000s of variables with ML
Relationships must be monotonic (move only in one direction)	Can model more realistic, non-monotonic relationships
Variables must be independent and be uncorrelated	No independence assumption made, no data constraints required
Results	
Status Quo Charge-Off and Approval Rates Unchanged Same Results	10-25% increase in approval rate, serving the underserved. Allows CU to price competitively in the mid-tiers
More sensitive to input fluctuation because fewer inputs are used	Models are more stable and less sensitive to changes in variables

Gain a clearer picture of every borrower



Incorporate more data
Often data you already have



Model more variables
Hundreds of variables and interactions
boost accuracy and resiliency

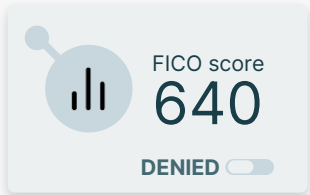


Swap risky borrowers for good ones
Custom scores and analyses boost performance

Current Process



- # accounts owned
- # late payments of 30-60 days
- # new credit accounts
- # years credit history
- # credit mix



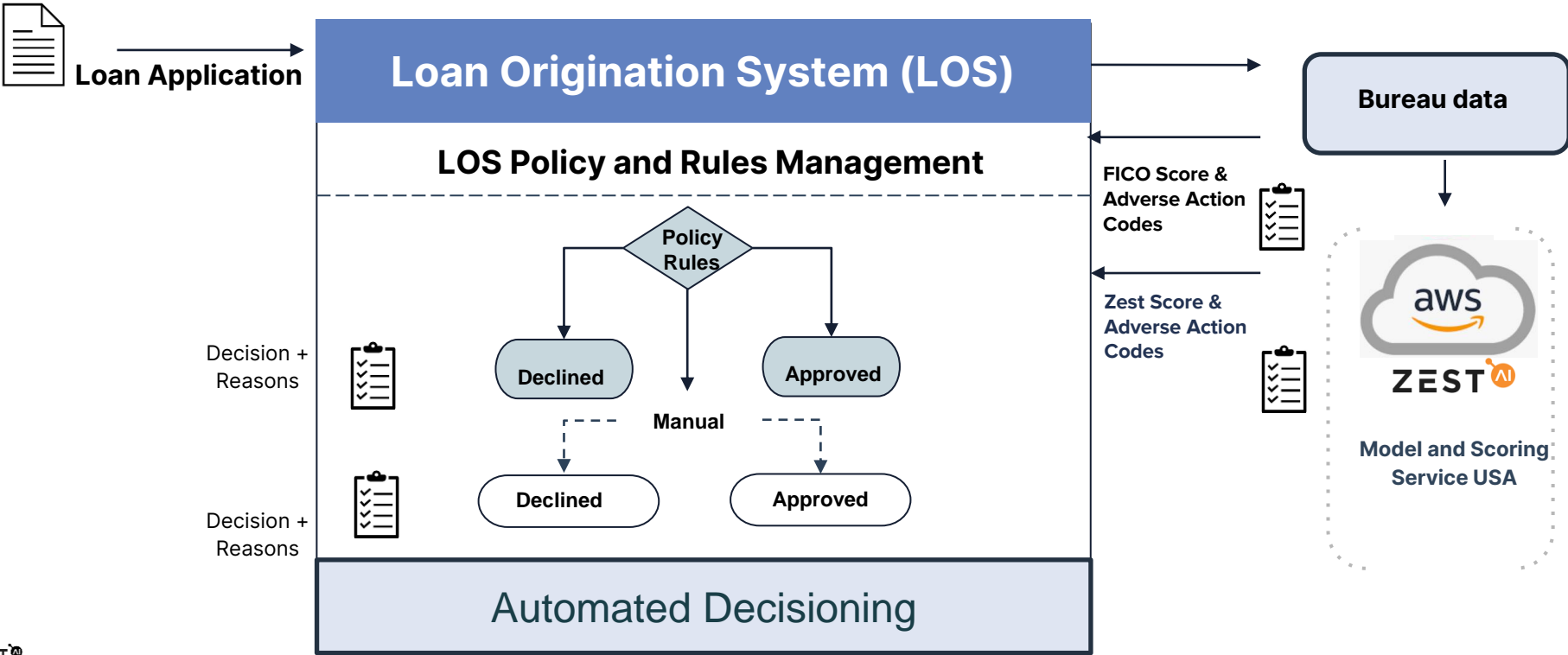
Automated Underwriting



- # accounts owned
- # late payments of 30-60 days
- # new credit accounts
- # years credit history
- # credit mix
- % of trades reported last month
- # of loans never delinquent
- Overall balance to credit amount
- Present status on all trades
- \$ total available credit limit
- Maximum single balance



Integrate Model with your LOS



Zest is Compliant

FCRA

Legislation passed in 1970 outlining rules for credit reporting and information management



Adverse Action Notices

This is important for model explainability and presents a challenge to many ML solutions. A model must be able to determine which key factors led to loan decision.

ECOA Reg B

Legislation passed in 1974 barring discrimination in lending based on protected classes



Fair Lending & Disparate Impact

This is important for model inputs and outcomes analyses to ensure a lender is not intentionally or inadvertently discriminating against a protected class.

SR 11-7

A supervisory letter from the Federal Reserve on proper development, implementation, and use of risk models in lending.



MRM Criteria & Guidelines

This requires extensive documentation regarding how a model was built, including validation exercises. It also includes how the model will be implemented and managed. Plus, compliance officers require separate fairness documents to ensure satisfactory model management.





Highly differentiated intellectual property, with underlying tools accepted by regulators and lenders



Zest IP assets

Fair Lending

- ✓ Automated disparate treatment and disparate impact testing
- ✓ Less discriminatory alternative search
- ✓ Adversarial de-biasing
- ✓ Adverse action

Explainability

- ✓ Human-readable descriptions explaining why model decided to approve/deny
- ✓ Feature importance & partial dependence plots
- ✓ Explaining stacked and heterogeneous models

Comprehensive, automated documentation



Regulator Acceptance

- Regulators deem Zest AI as the “**Gold Standard**” in ML
- Zest AI clients’ have **passed risk management audits** with Federal regulators
- Zest AI is a **supervised service-provider to covered entities**, and is under the purview of the CFPB
- Zest AI has trained top bank examiners at FDIC, OCC, Fed Reserve, CFPB, and FHFA on how to **examine ML models**

Proof of Concept Model Results

ABC Credit Union



Data sheet to build preliminary model

Product Line Information DirectAuto	Portfolio Level Statistics	Credit Tier Stratification (Use your actual credit tiers)							
		A+	A	B	C	D	E		
	Portfolio Total	Super prime	Prime	Prime	Near Prime	Subprime	Subprime		
FICO/Vantage Score Range		720+	690-719	660-689	640-659/No score	600-639	<=599		
Total Annual Applications (#)	1774	1035	200	150	148	96	145		
Automated Decision %	0								
Total Annual Approved (#)	1299	877	169	106	68	54	46		
Total Annual Funded (#)	1276	867	139	106	66	51	45		
Total Annual Origination (\$)	\$ 27,455,511.39	\$ 20,225,509.27	\$ 2,926,577.71	\$ 1,923,112.39	\$ 991,069.75	\$ 747,754.44	\$ 641,487.83		
Average Funded Amount	\$ 21,516.85	\$ 23,247.21	\$ 21,054.52	\$ 18,142.57	\$ 15,016.21	\$ 14,661.85	\$ 14,579.27		
Average Interest rate charged	3.52	2.37	3.51	5.1	6.37	9.67	10.94		
Average Loan Term	63								
Average Loan life	24								

Product Line Information Indirect Auto	Portfolio Level Statistics	Credit Tier Stratification (Use your actual credit tiers)							
		A+	A	B	C	D	E		
		Super prime	Prime	Prime	Near Prime	Subprime	Subprime		
FICO/Vantage Score Range		720+	690-719	660-689	640-659/No score	600-639	<=599		
Total Annual Applications (#)	3369	966	315	381	470	466	771		
Automated Decision %	0								
Total Annual Approved (#)	1150	670	142	130	92	79	37		
Total Annual Funded (#)	710	336	93	80	76	63	62		
Total Annual Origination (\$)	\$ 18,146,720.00	\$ 8,869,221.00	\$ 2,678,208.00	\$ 2,146,759.00	\$ 1,585,391.00	\$ 1,493,228.00	\$ 1,373,912.00		
Average Funded Amount	\$ 25,416.00	\$ 26,396.00	\$ 28,798.00	\$ 26,834.00	\$ 22,051.00	\$ 23,702.00	\$ 22,160.00		
Average Interest rate charged	5.25	2.97	3.99	5.46	7.14	10.53	11.72		
Average Loan Term	67								
Average Loan life	29								

ABC CU's Automation Goals

- Saying “Yes” to more members
- More fairness to all members regardless of class
- Keeping risk constant in targeted lower tiers
- Compliant for all applicants

ABC CU Economic Analysis

Increase Approvals Safely



Using more data and better math

Results for ABC Credit Union



Model Inputs

Bureau-only model trained using over 1 million borrowers

Modeling Target

Bad = DQ60+ in 24 months

*DQ60+ in 24M shows a **high true positive** rate and a **low false positive** rate versus charge-off. Longer seasoning window less of a concern given large sample size.*

Bureau Tables

- 1 Tradelines
- 2 Inquiries

Tens of millions of individual data points across these two bureau tables provide the model ample information from which to learn repayment patterns.

Train & Test Splits

SET	TIME PERIOD
Train	January 2016 - June 2018
Test	July 2018 - December 2018

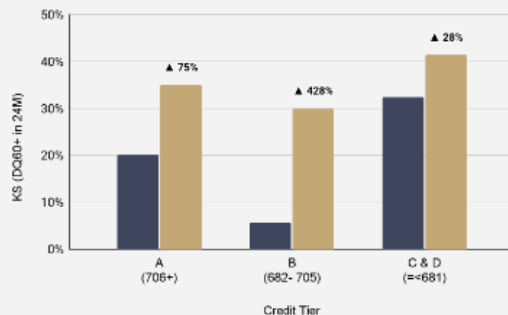
*Millions of borrowers in Training represents **a massive improvement** over Client data alone. Test set provides unbiased view of model lift and stability.*

Statistical Accuracy

KS comparison indicates massive lift over the National Credit Score across credit tiers

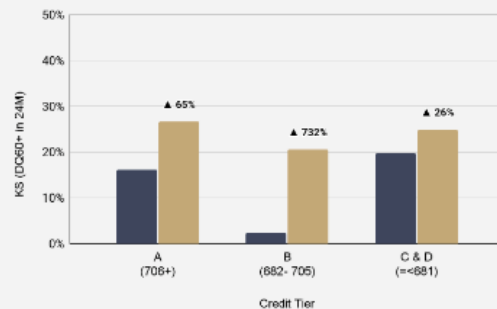
AUTO

Zest vs. National Credit Score



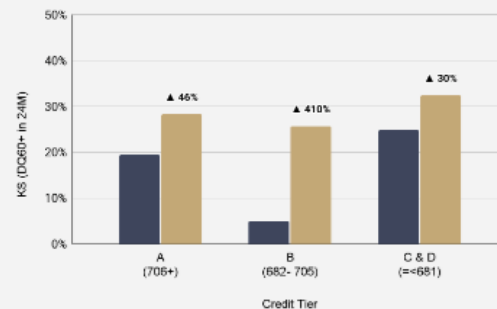
PERSONAL LOAN

Zest vs. National Credit Score



CREDIT CARD

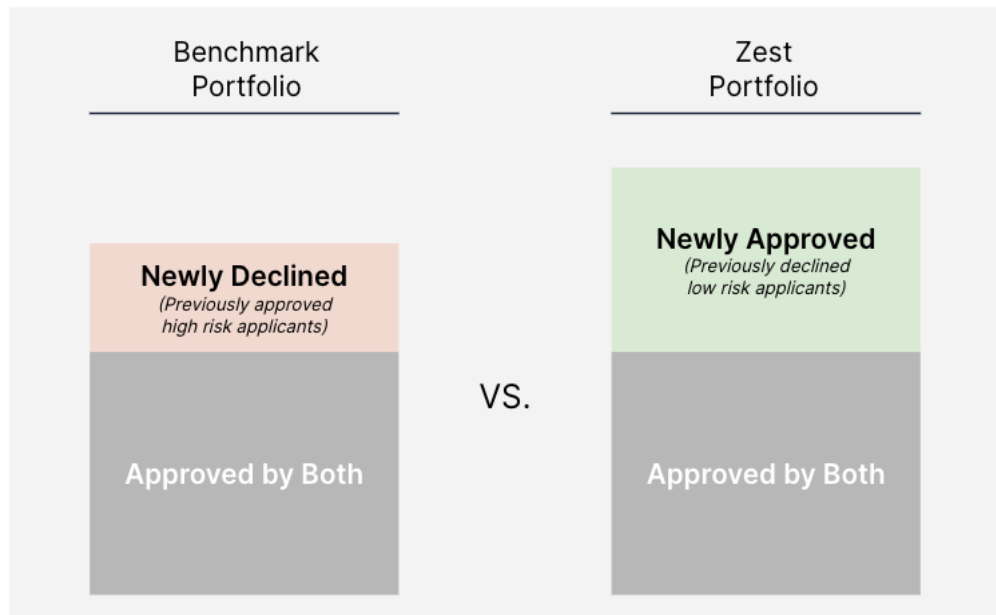
Zest vs. National Credit Score



LEGEND ● National Credit Score ● Zest

Approve more good borrowers

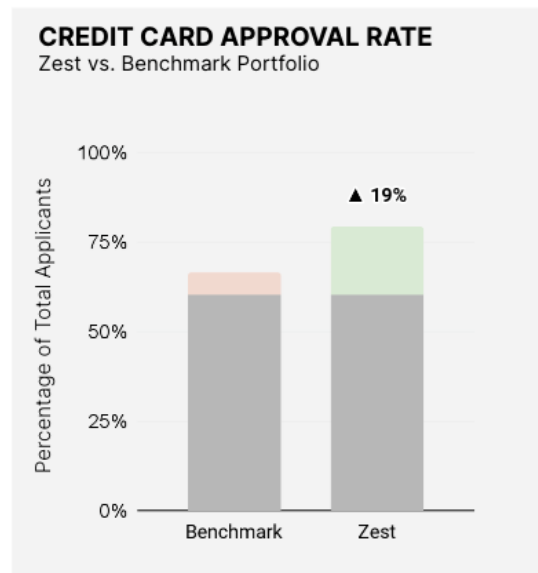
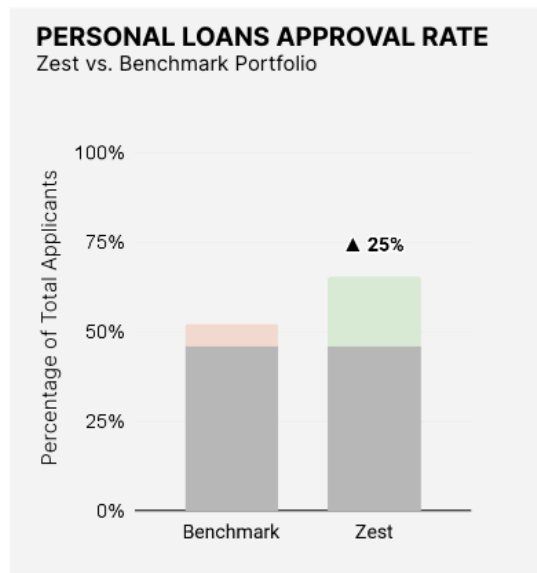
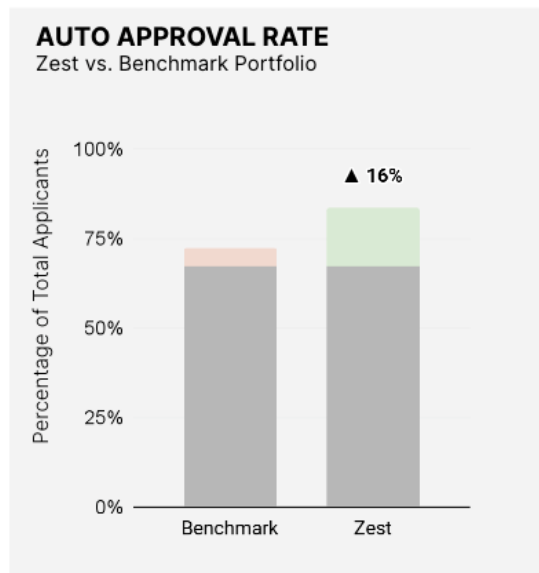
Economic value is estimated by comparing the size of the Benchmark portfolio to the Zest portfolio



By replacing high risk borrowers with low risk ones, you can approve significantly more borrowers without increasing risk.

All Products: Approvals

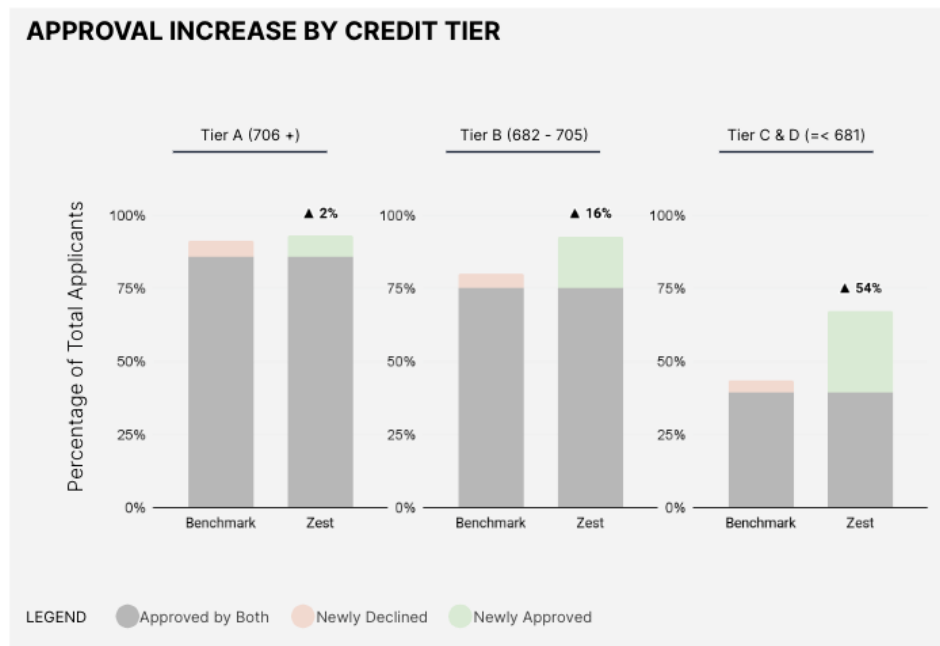
Significant approvals across product portfolios while holding benchmark risk constant



LEGEND Approved by Both Newly Declined Newly Approved

Auto: Increase approvals by 16%

Holding risk constant, we calculate the auto approval rate increase per tier enabled by the Zest model

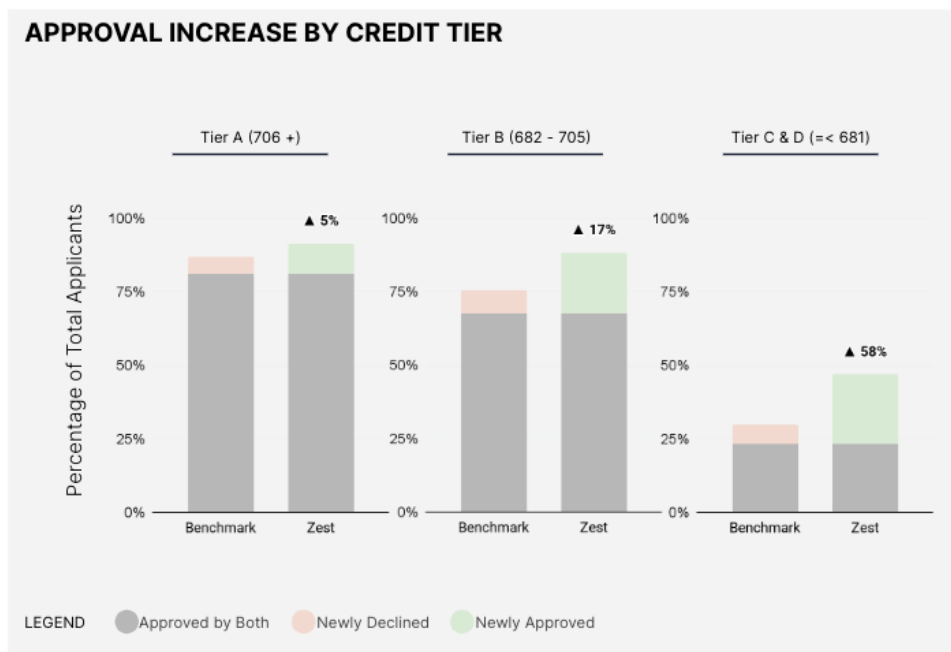


16%

→ Overall approval rate increase versus Benchmark

Personal: Increase approvals by 25%

Holding risk constant, we calculate the auto approval rate increase per tier enabled by the Zest model

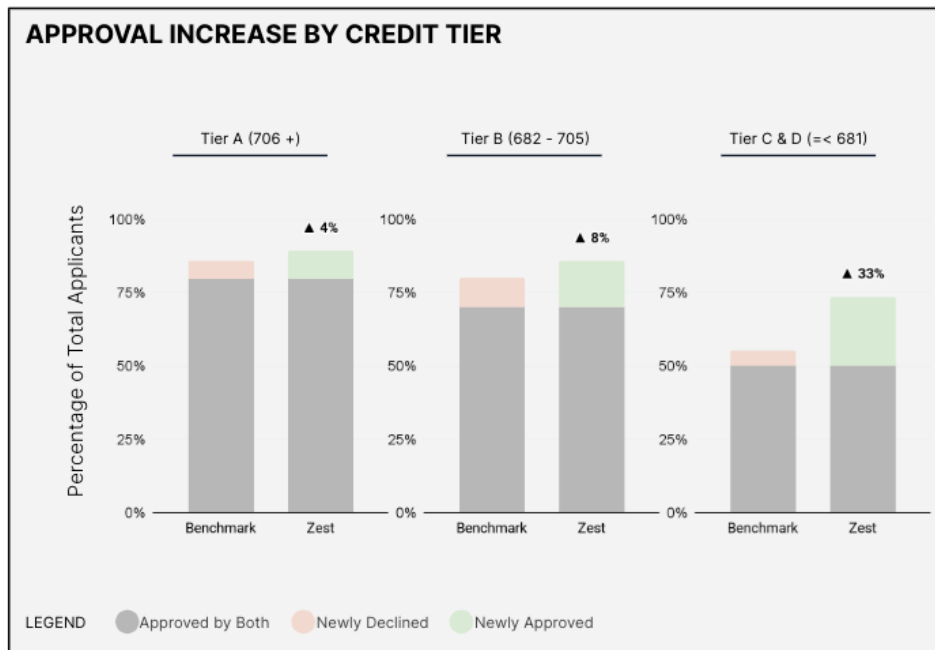


25%

→ Overall approval rate increase versus Benchmark

Credit Card: Increase approvals by 19%

Holding risk constant, we calculate the auto approval rate increase per tier enabled by the Zest model

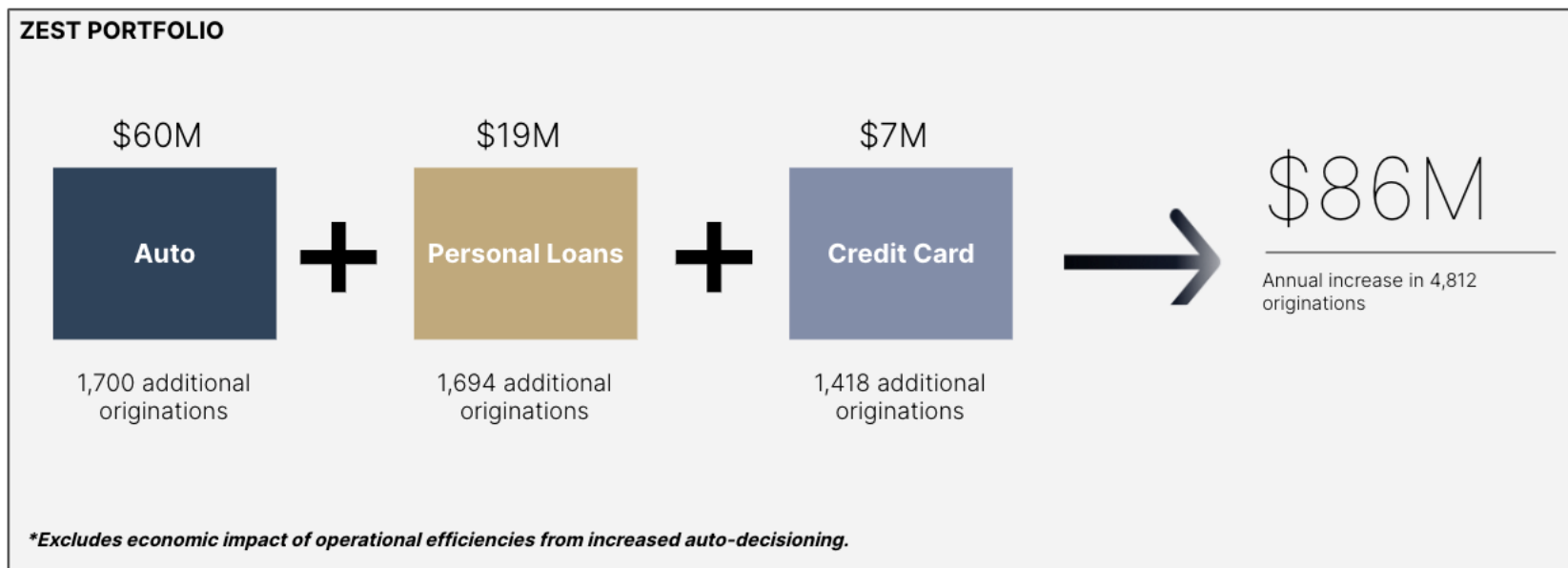


19%

→ Overall approval rate increase versus Benchmark

ABC CU Says "Yes!" to more members

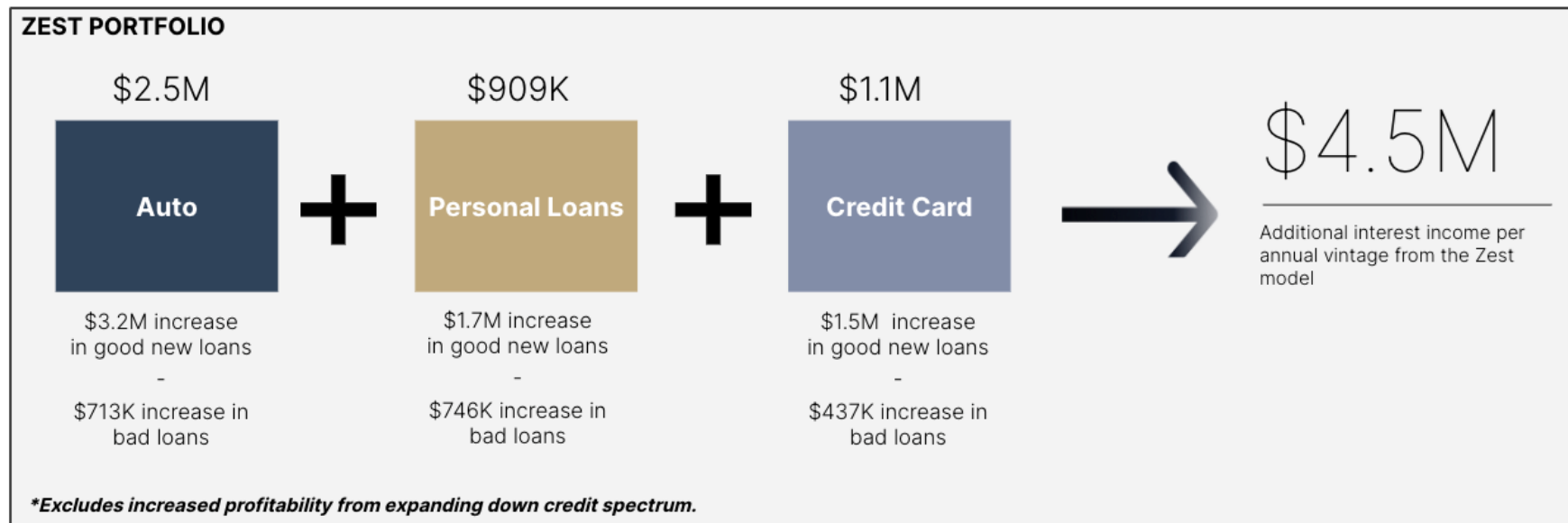
Approval increases across product portfolios equates to 4,812 more loans without increasing risk



Note: Applications figure based on EECU's data. Booking rates also assumed to be held constant.

ABC Credit Union's Economic impact

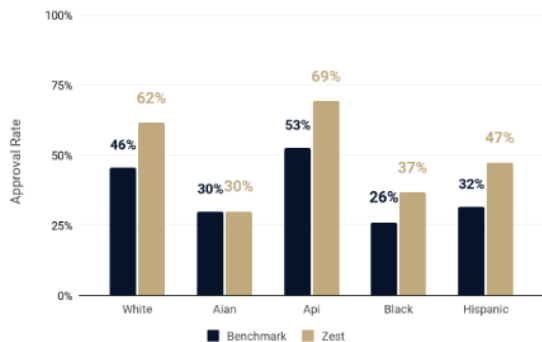
We then measure the total economic impact of the approval increase in the auto, personal loans, and credit card portfolios



Automation increases ABC CU's approval rates for protected borrowers – with no increase in risk

Zest will increase Hispanic & Black approvals by +40% and Female approvals by 36%

Race/Ethnicity: Approval Comparison

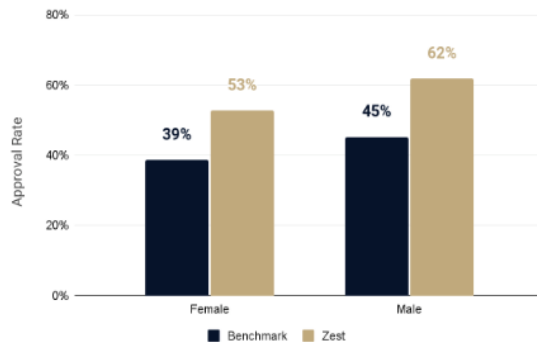


49% Increase in Hispanic approvals

41% Increase in Black approvals

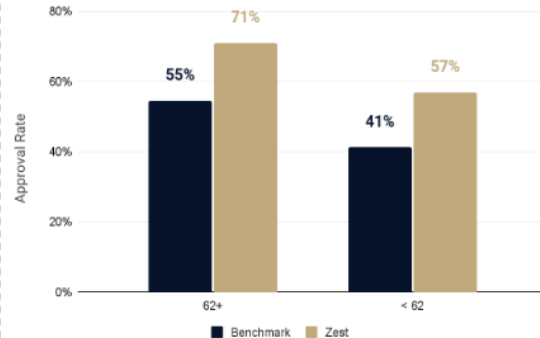
31% Increase in AAPI approvals

Gender: Approval Comparison



36% Increase in female approvals

Age: Approval Comparison



30% Increase in elderly approvals

POC Evaluation Process

1. Provide Zest with data sheet (1-3 days)
2. Zest builds auto decisioning model (30-days)
3. Present CU specific model to stakeholders (1 day)
4. If CU proceeds, we then enhance model with member specific data and historical [5 years] (30 days)
5. Integrate to LOS (2-15 days)

Summary



Evaluating Auto Decisioning Solutions

- **Compliance**- CFPB, ECOA, SR 11-7, FCRA
- **PII**- Protection of your members PII of being accessed off-shore or unknown by 3rd parties
- **Data Consultants**- Access to data consultants
- **LOS Integration**- Partnership Agreements with LOS vendors or authorization to access LOS via API

Summary

- AI and ML are not robots, but the processing of more data with better math
- Implementing auto decisioning is NOT disruptive unlike a LOS or Core conversion
- Auto decisioning should be your CU's highest priority in 2022 so you can say yes to more members in 2023 and be well positioned for the uncertain times ahead

Questions?

Contact Craig Peterson
cmp@zest.ai



Thank you



More approvals, lower risk,
happier members

