



# *High Cost Claim Prediction for Actuarial Applications*

Vincent Kane, FSA, MAAA

Research Scientist, DxCG- A Division of Urix Inc.

The Second National Predictive Modeling Summit

Washington, D.C.

September 22, 2008

## Predictive Modeling vs. Risk Adjustment

- PM: Predict claims \$ or stratify risk for people or groups, by any means necessary
  - Uses detailed claim-based diagnosis information and possibly procedure data, utilization data, prior costs, timing of claims, benefit provisions, lifestyle-based variables or HRA data, credit info, kitchen sink
- RA: Quantify differences in health status among populations and over time to discover illness burden
  - Picks up on differences in health status and health status alone. Risk *assessment* characterizes the relative cost differences for persons or groups, for example, using relative risk factors.

## Choice of a predictive model versus risk adjuster

- If risk-adjusting payments to providers or plans, you may not want to include prior utilization, costs or procedures.
  - Fairly assess health status, therefore, ignore diagnosis codes that are vague, difficult to audit, and gameable.
- For underwriting, care management, and stop loss or reinsurance applications, you may want to use all available predictors
  - Could recalibrate standard risk adjustment models by adding new variables, or
  - Build a predictive model from scratch for the intended application

## “High Cost Case Model” (HCCM)

- A predictive model which uses all diagnoses and pharmacy claims to prospectively find members likely to be high cost
- Based on RxGroups® and HCC clinical groupings
  - Adds proprietary variables based on prior year cost and utilization patterns
    - Blood disorders, cancers, CHF, diabetes, usual suspects
    - Extremely high cost drugs, certain injectables, etc.
  - Assumes fully run out claims
  - Does not use a lag before the prediction period

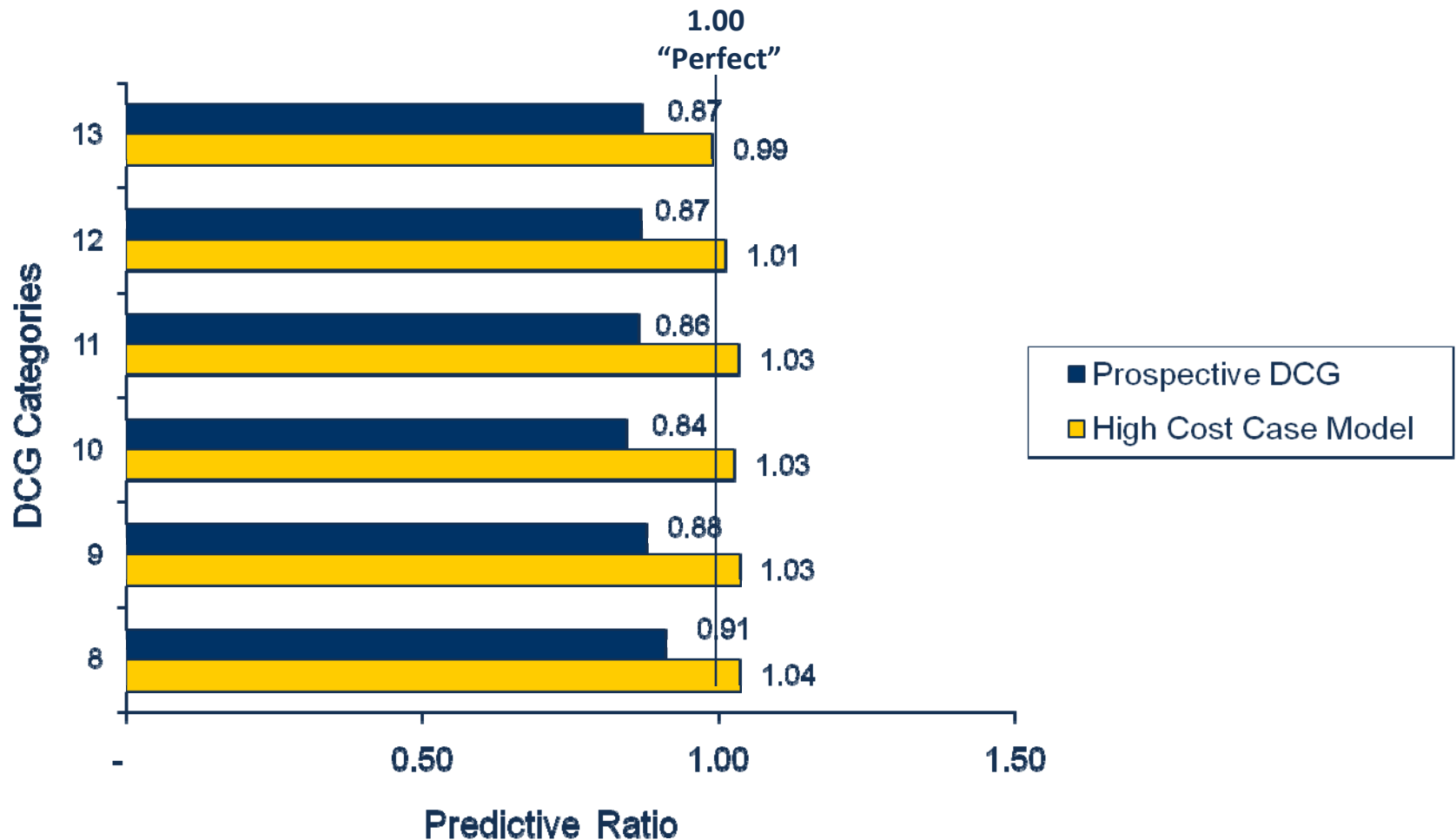
## HCCM - Model Characteristics

- Calibrated w/ Thomson MedStat Marketscan data
- Dependent variable, and therefore outcome to be predicted, are year 2 total allowable claims costs
  - A year 2 risk score is the model output
- Prospective with top coding choices
  - No top coding
  - Top coded at \$250k
  - Top coded at \$100k
  - Top coded at \$25k

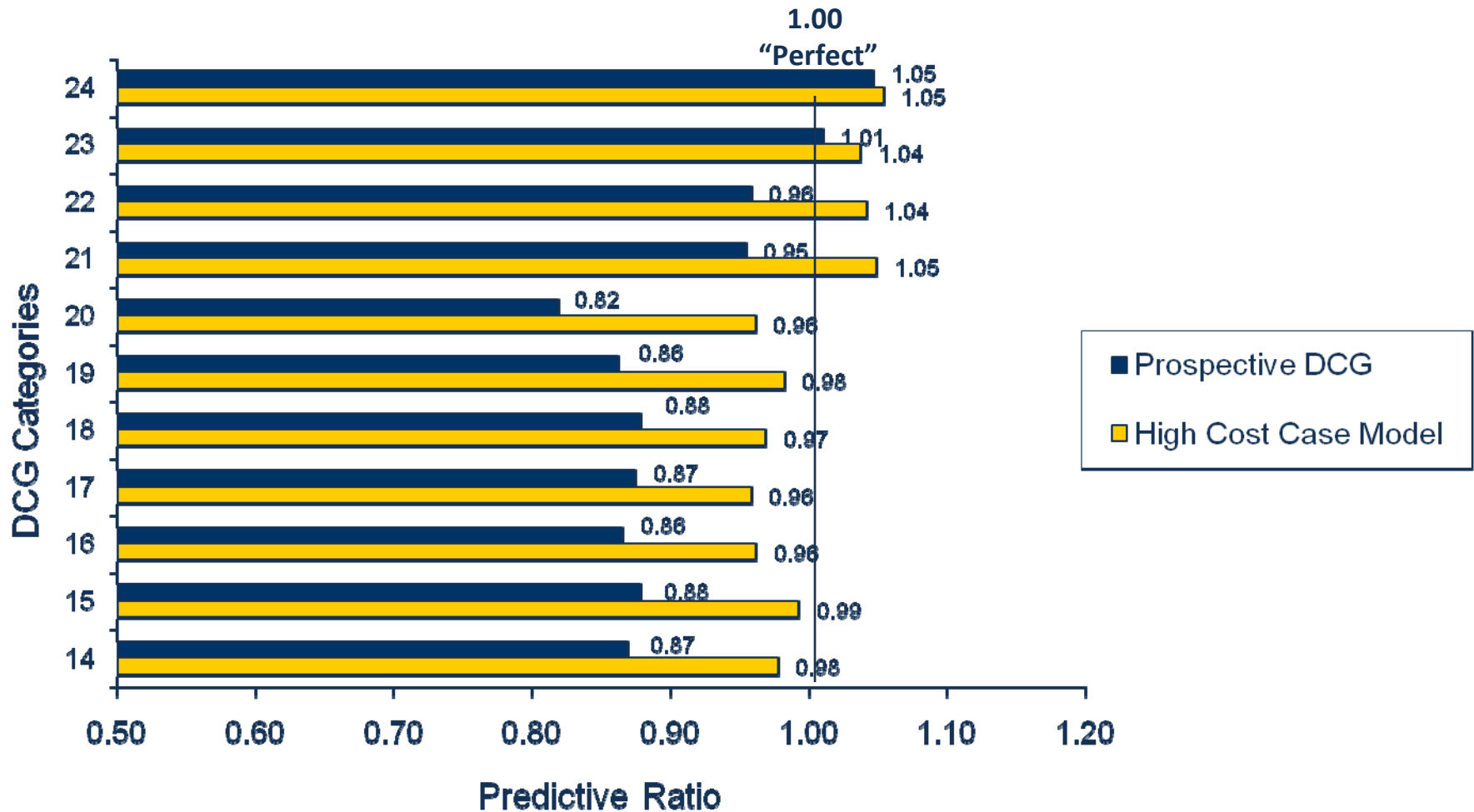
## How is HCCM Different From Prospective DCG/HCC Model?

- Uses prior costs and RxGroups® (NDC codes) as inputs
- Higher R-squared (22.1% vs 14.1%)
- Improved predictive ratios
- Performs better in top ½% and 1%
- Has a higher Positive Predictive Value (PPV) for predicting high cost patients

## HCCM Performs Better In Low DCG Buckets and ...

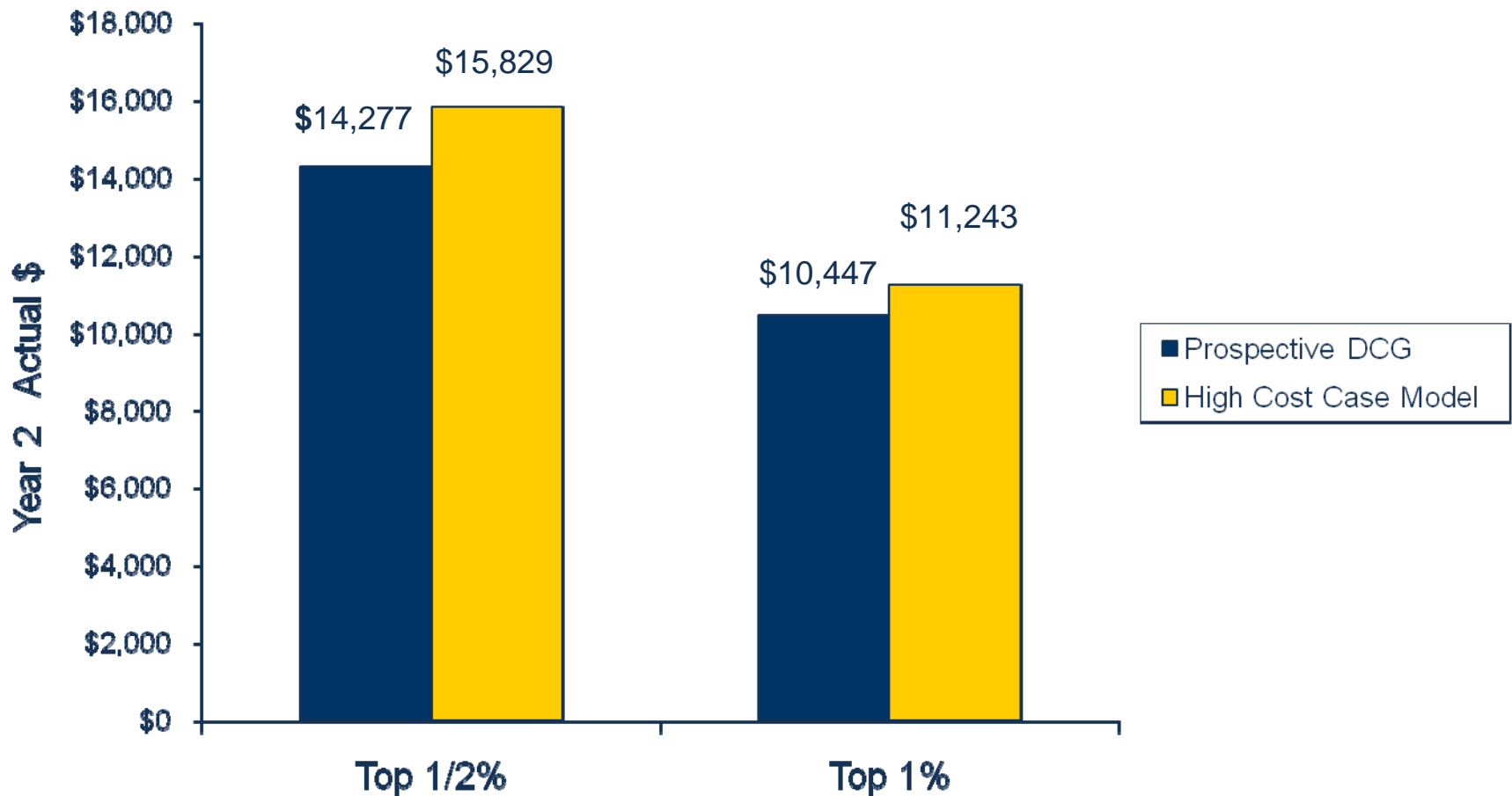


## ...Performs Much Better In High DCG Buckets

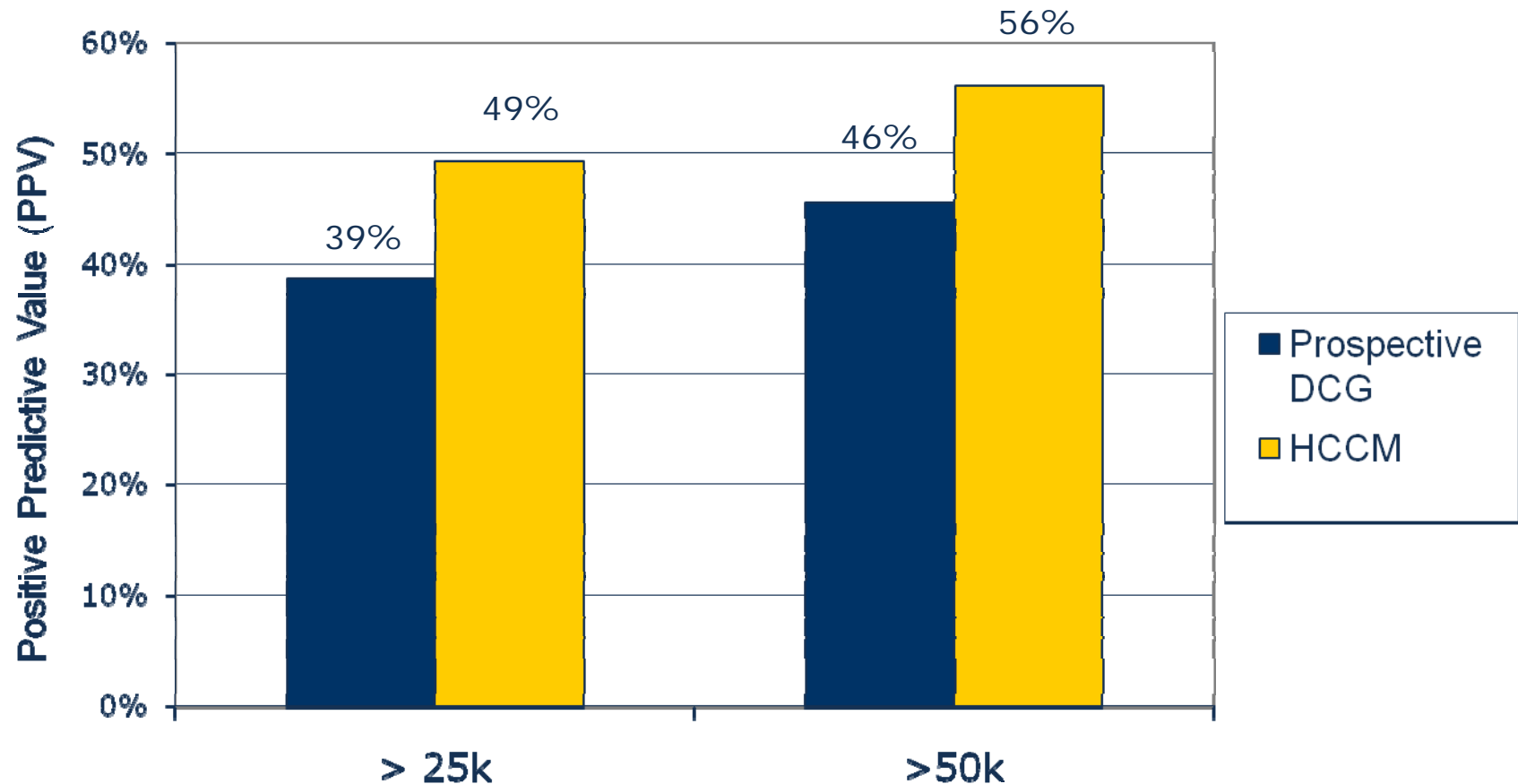




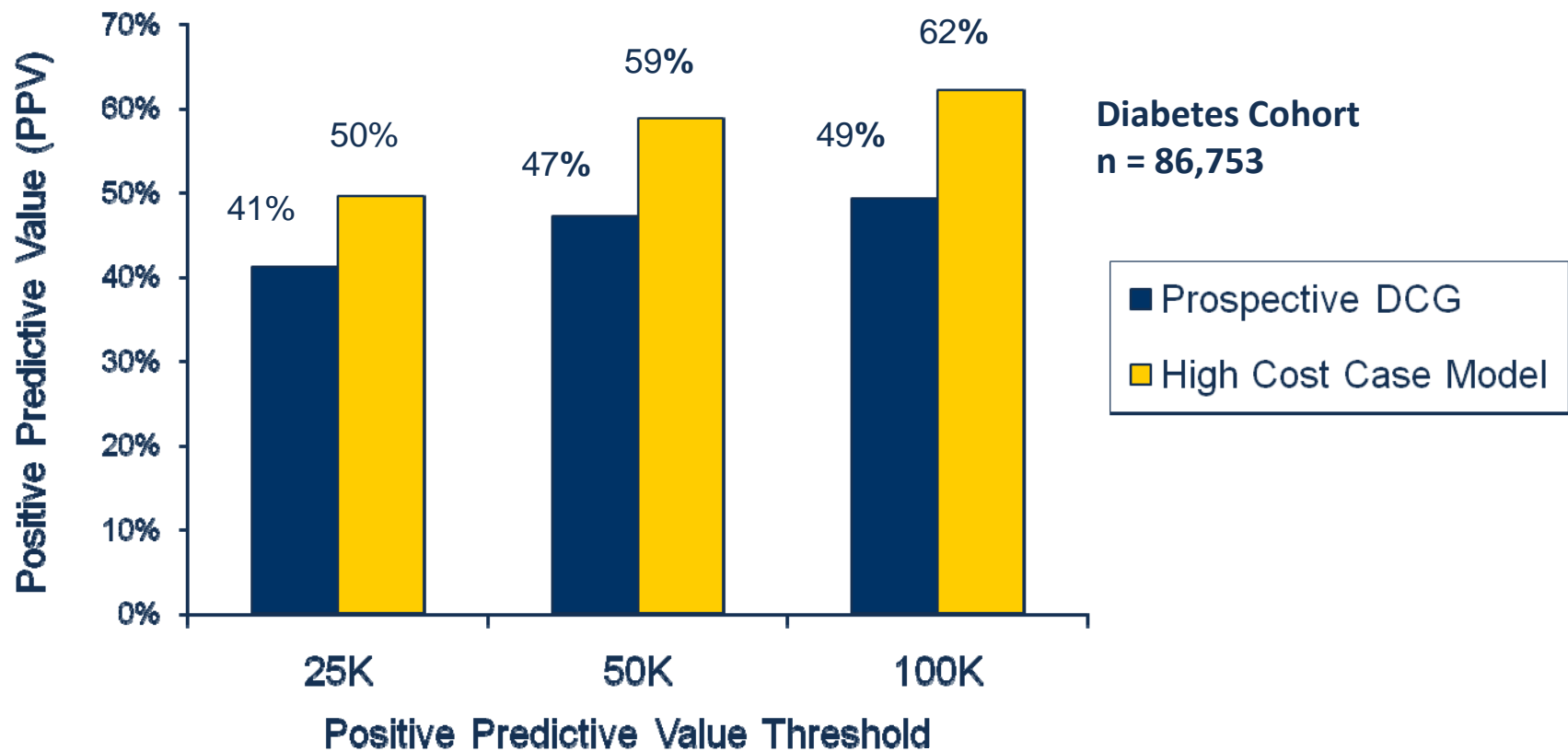
## HCCM Finds More Expensive Individuals in Top Groups



## HCCM Correctly Predicts More Expensive Individuals



## HCCM Correctly “Finds” More Cases – *PPV for Diabetic Cohort*



## Comparing HCCM with Other Means of Predicting Future Costs

- There are lots of different approaches that may be used to predict future costs
  - Age-sex
  - Prior year cost
  - Prospective DCG model
  - Prospective RxGroups model
  - Parametric methods using distributional forms
  - Two-part models
  - Other econometric models
  - Data mining techniques
  - *Combinations of methods*

*Upgrading* the standard DCG-HCC model to create one type of “Combined Method”

- In the MarketScan database, DxCG created a model to simulate the combination of the traditional methods
- The recalibration combines age sex categories, the prospective DCG score and year 1 costs to predict year 2 costs
- We define this as the “Combined Method”

## “Predictive Model” performance versus standard diagnosis-based risk adjusters

	R-Squared
Prospective DCG	14.1%
Combined Method (Prospective DCG and Prior Costs)	16.5%
HCCM (no top coding)	22.1%

Predictive performance improves with decreasing top-coding thresholds

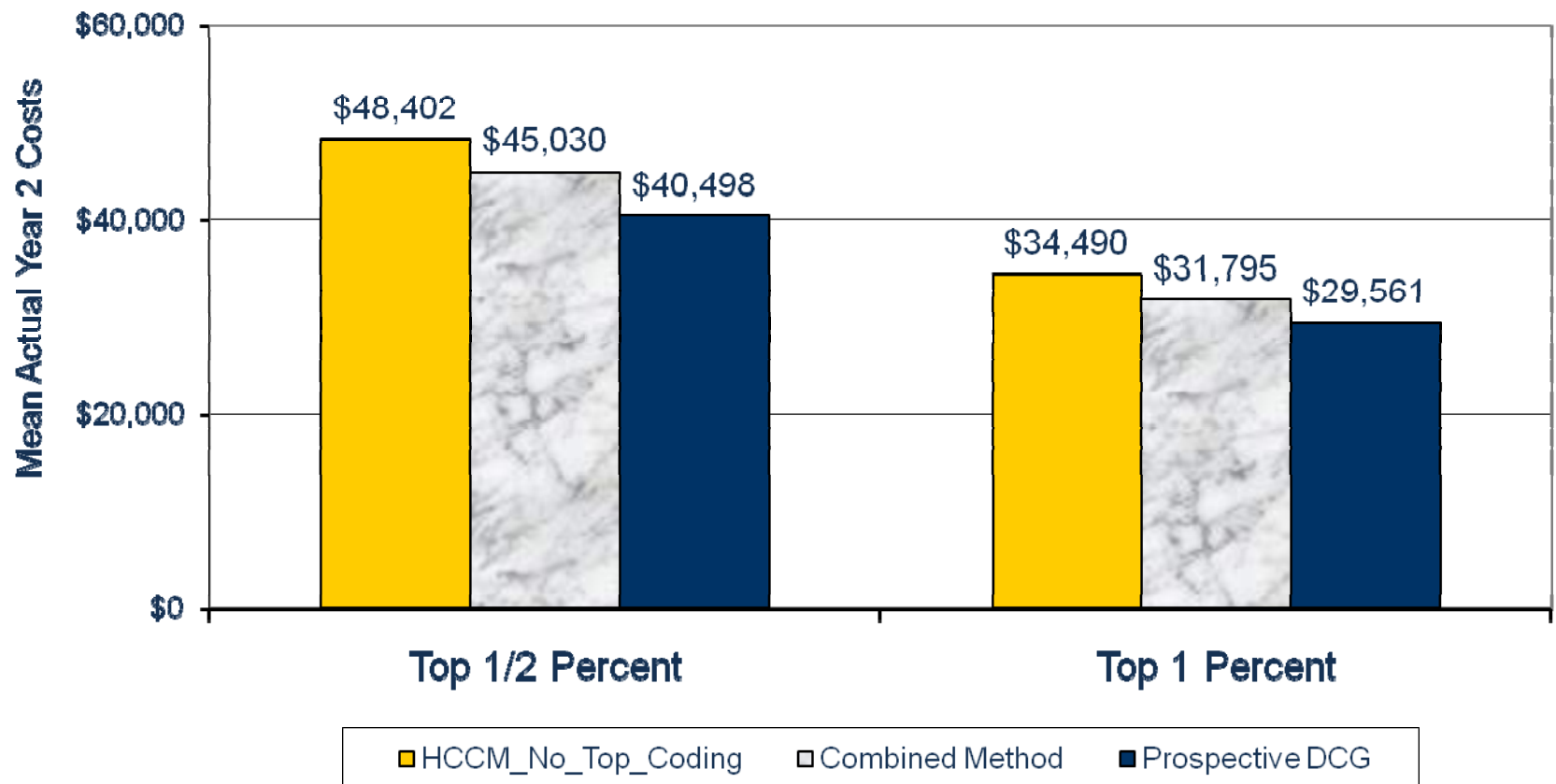
High Cost Case Model	R-squared
No Top Coding	22.1%
\$ 250k	26.6%
\$ 100K	28.8%
\$ 25K	31.4%

## Also possible to create “top groups” for each model

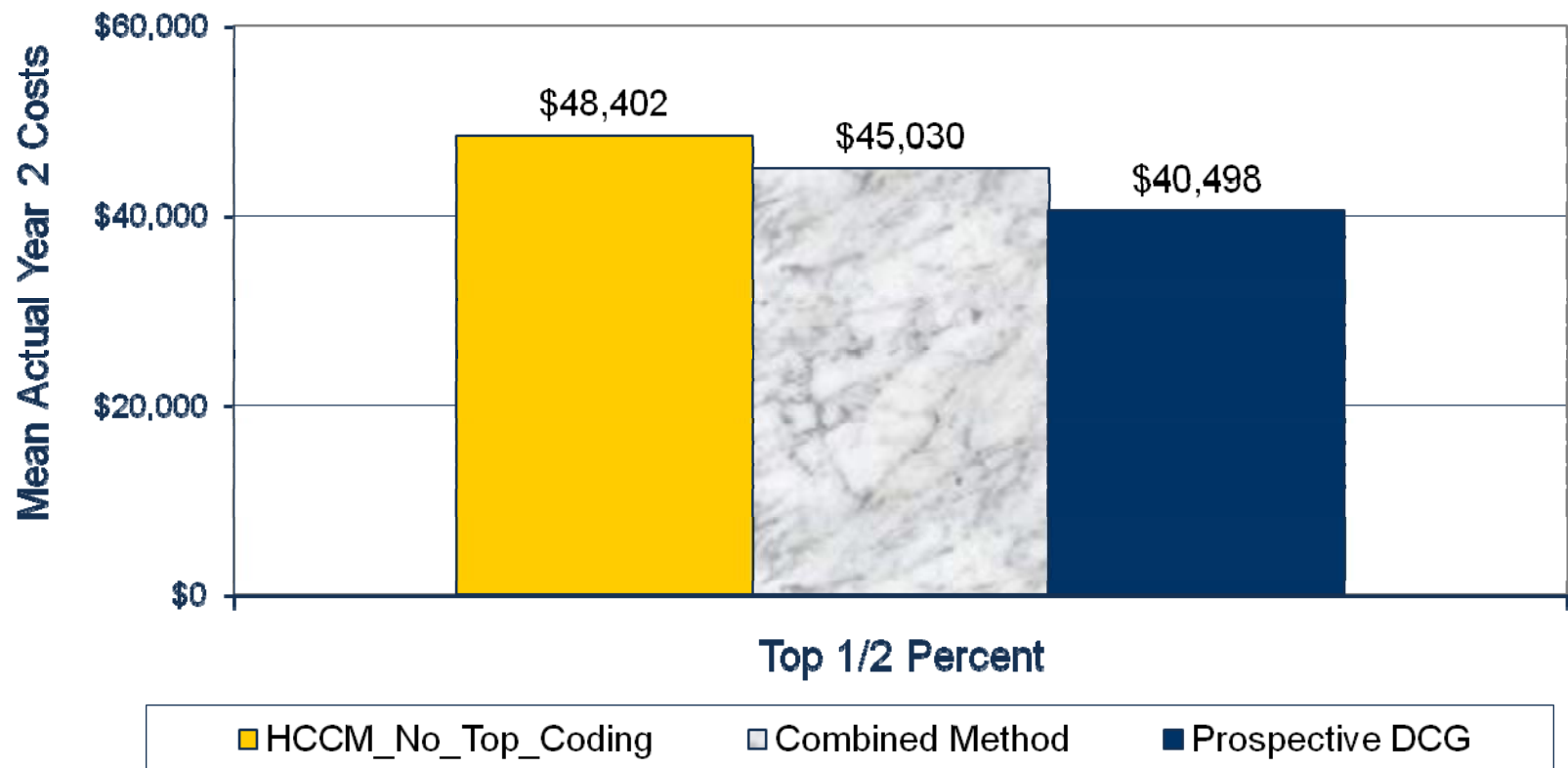
- Top groups using the prospective DCG model
  - Members who were in the top ½ percent using the prospective DCG method (N= 12,727)
  - Members who were in the top 1 percent using the prospective DCG method (N= 25,453)
- Top groups using the combined method
  - Members who were in the top ½ percent using the combined method (N= 12,727)
  - Members who were in the top 1 percent using the combined method (N= 25,453)
- Top groups using HCCM (no top coding)
  - Members who were in the top ½ percent using HCCM (N= 12,727)
  - Members who were in the top 1 percent using HCCM (N= 25,453)



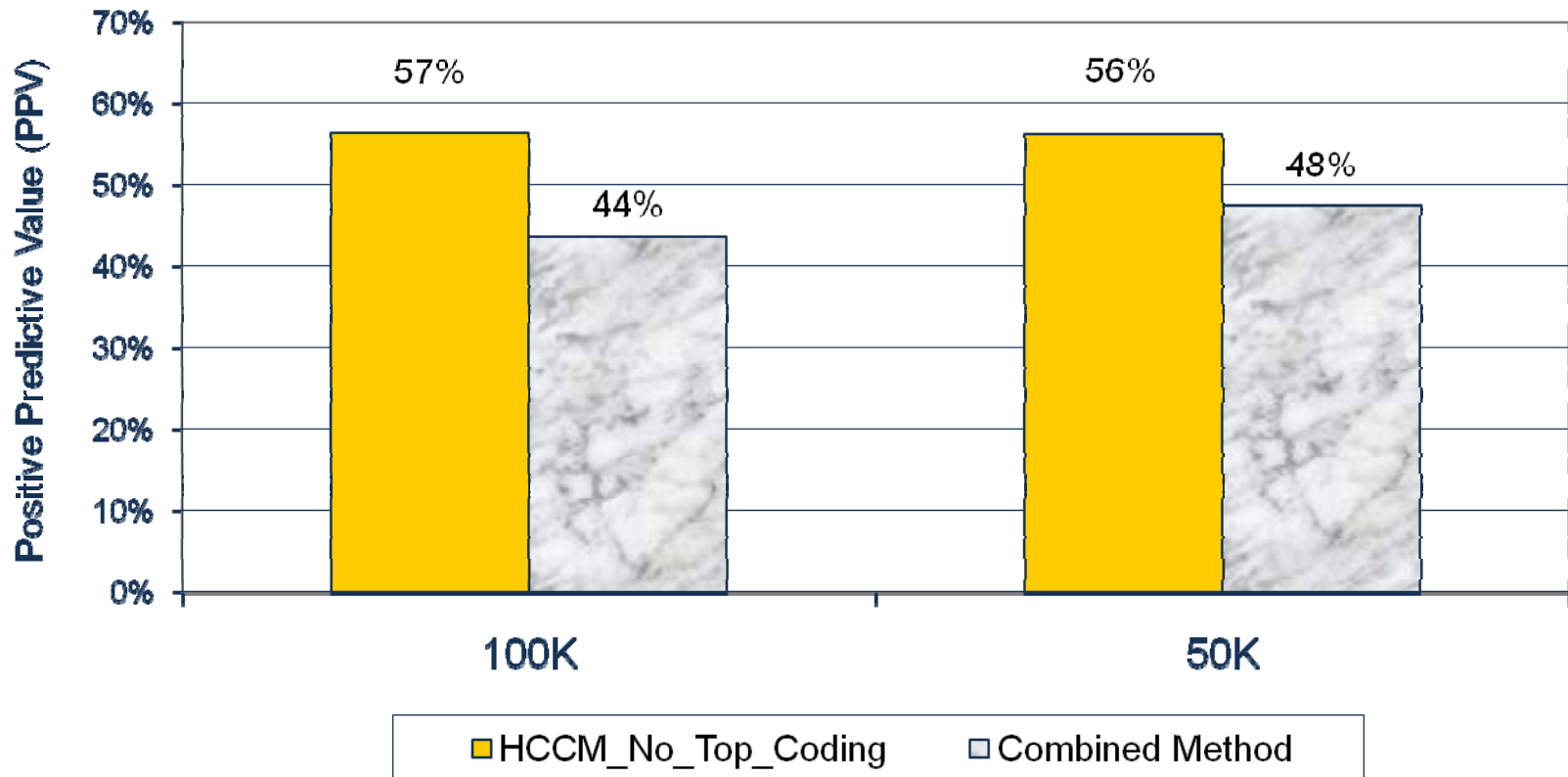
# HCCM Identifies Members With Higher Average Actual Year 2 Costs



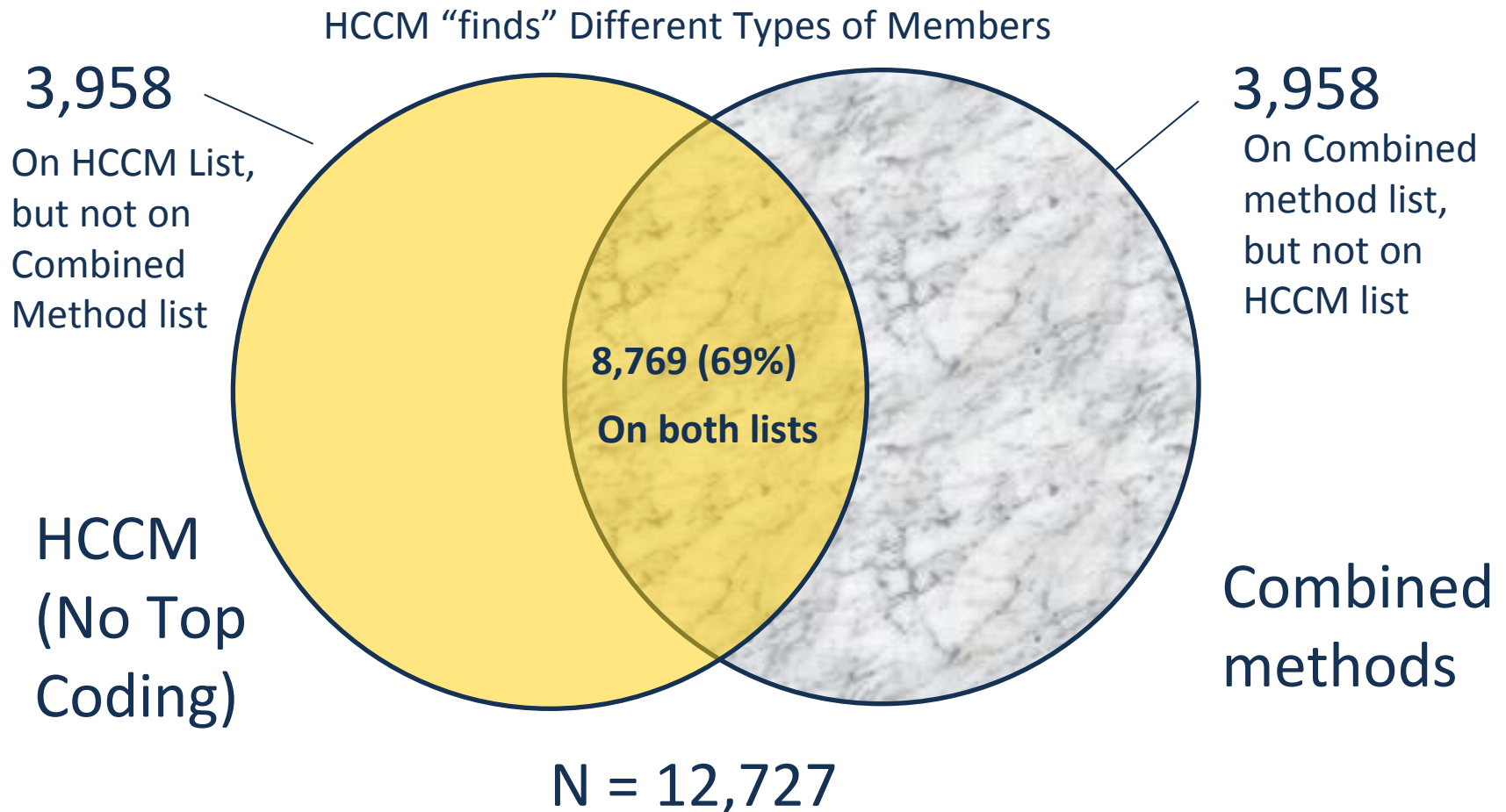
## Results for the top ½ percent group (N = 12,727)



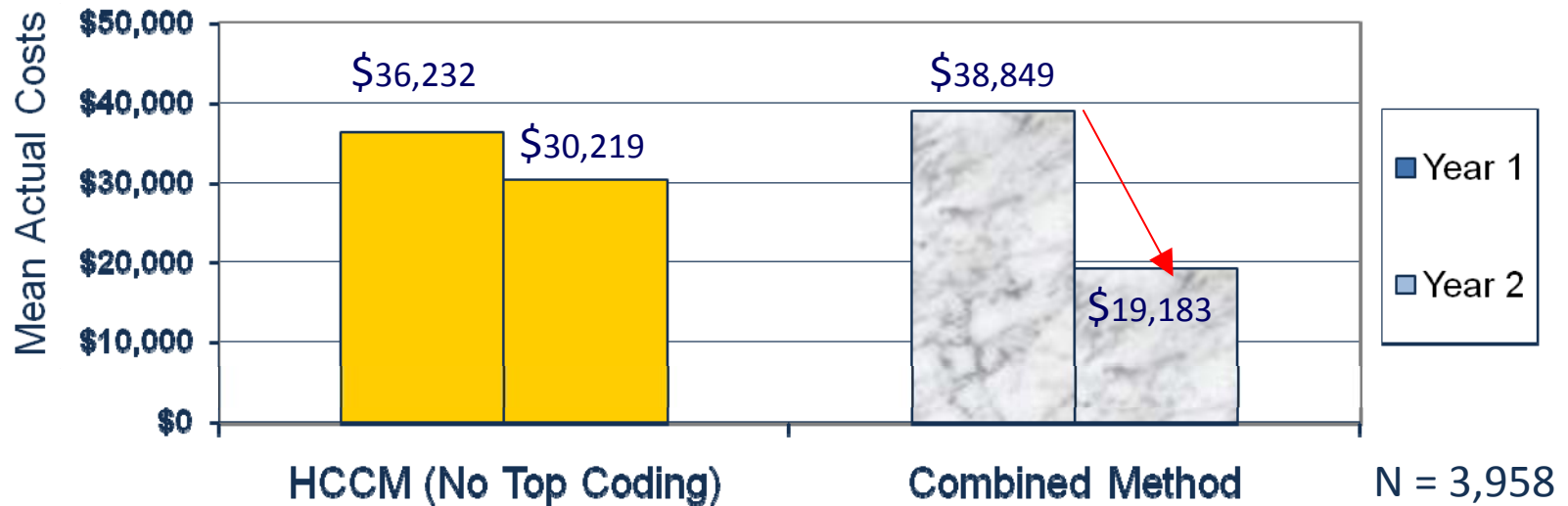
## HCCM Has a Higher PPV Compared to the Combined Method (N = 12,727)



# HCCM Model Found 3,958 Individuals Not On the List from the Combined Method



## The 3,958 Non Overlapping Members Identified by the Combined Method Illustrate Regression To The Mean

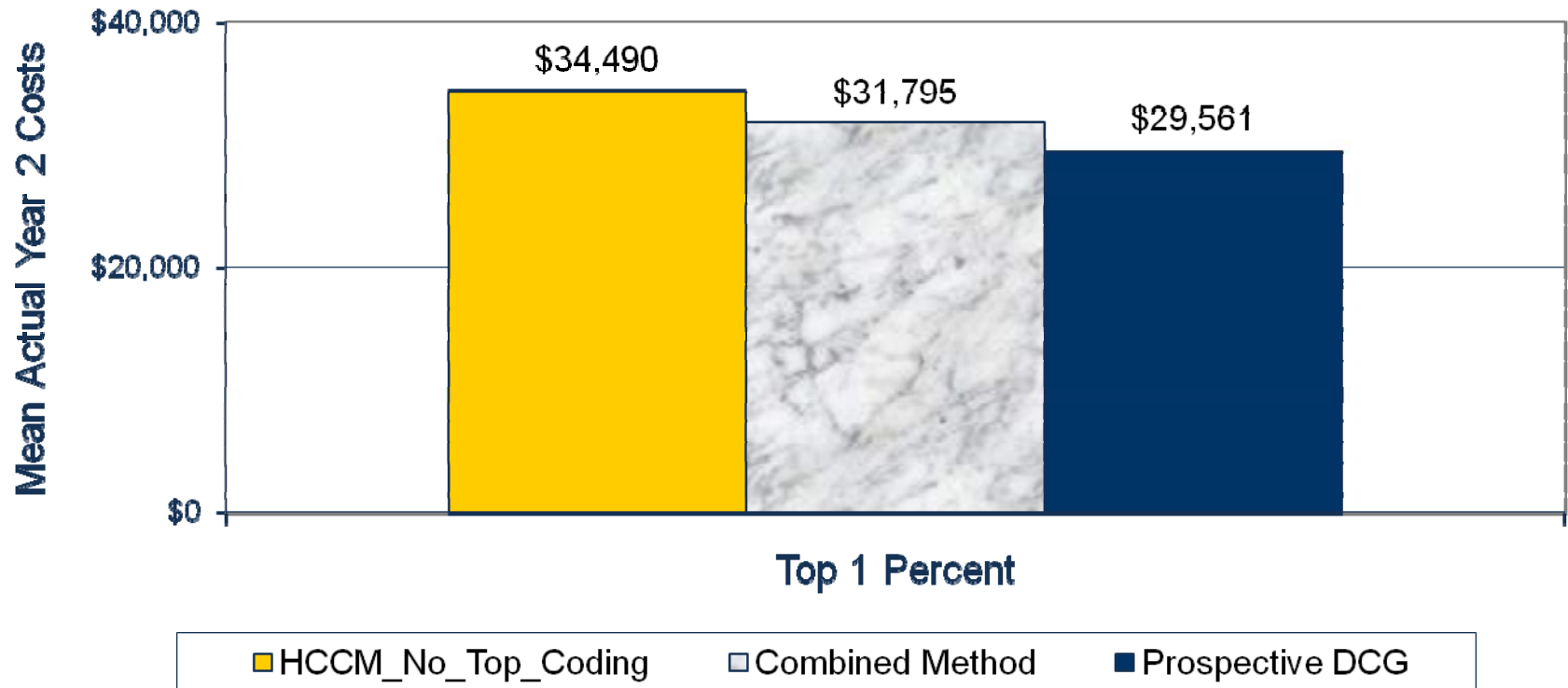


Costs for the Non Overlapping 3,958 Individuals on the Combined List drop by 51% in Year 2. By contrast, the non overlapping 3,958 Individuals on the HCCM List drop by only 17% in Year 2

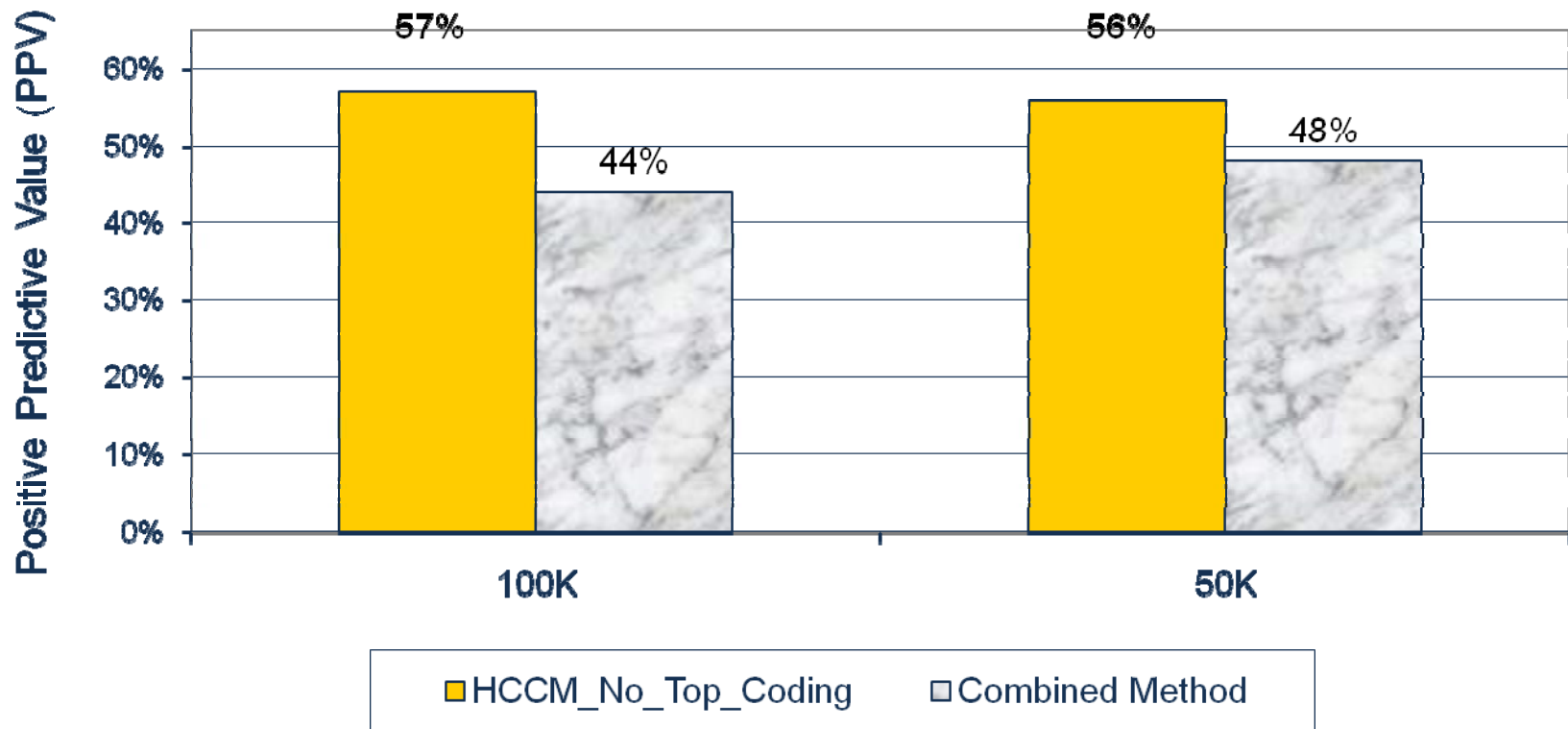
## The HCCM Model Identifies High Cost Cases Better than Traditional Methods

- 3,958 non overlapping individuals on the HCCM list had total Year 2 costs of more than \$120 million
  - Average PMPY is \$30,219 as shown on the previous chart
- 3,958 non overlapping individuals on the Combined method list had total Year 2 costs of \$76 million
  - Average PMPY is \$19,183 as shown on the previous chart

## Results for the top 1 percent group (N=25,453)



## HCCM Has a Higher PPV Compared to the Combined Method (N = 25,453)





# HCCM Model Found 8,390 Individuals Not On the List from the Combined Method

HCCM “finds” Different Types of Members

**8,390**  
On HCCM List,  
but not on  
Combined  
Method list

**8,390**  
On Combined  
method list,  
but not on  
HCCM list

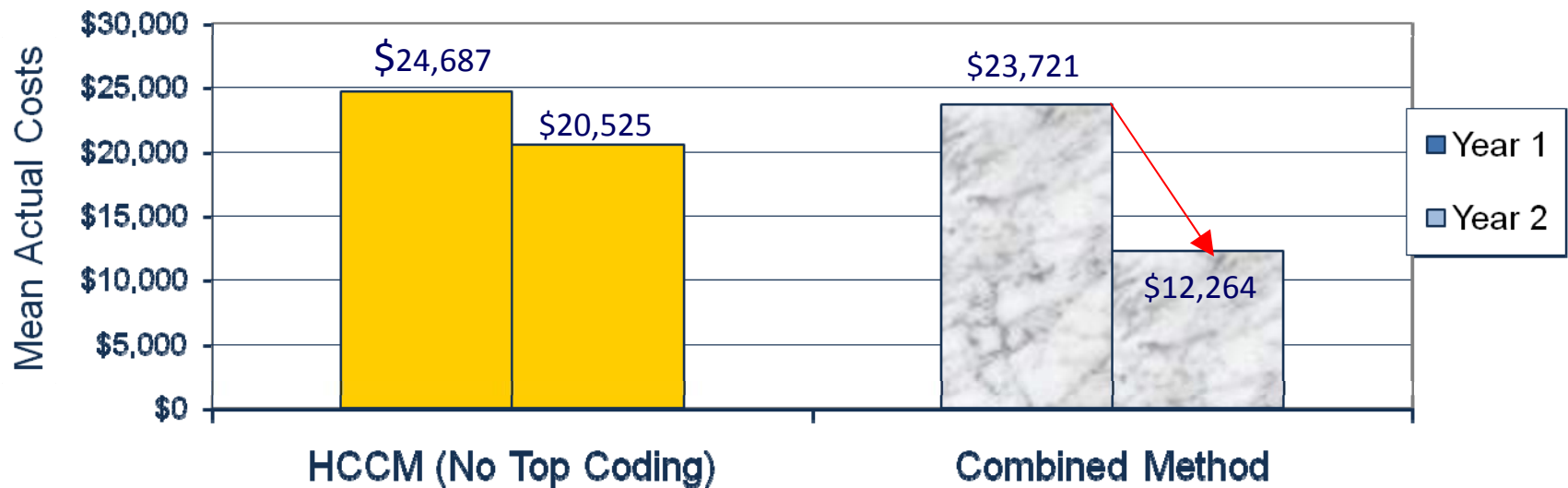
**17,063 (67%)**  
On both lists

HCCM  
(No Top  
Coding)

Combined  
methods

**N = 25,453**

## The 8,390 Non Overlapping Members Identified by the Combined Method Illustrate Regression To The Mean



Costs for the Non Overlapping 8,390 Individuals on the Combined List drop by 48% in Year 2. By contrast, the non overlapping 8,390 Individuals on the HCCM List drop by only 17% in Year 2

## The HCCM Model Identifies High Cost Cases Better than Traditional Methods

- 8,390 non overlapping individuals on the HCCM list had total Year 2 costs of more than \$172 million
  - Average PMPY is \$20,525 as shown on the previous chart
- 8,390 non overlapping individuals on the Combined method list had total Year 2 costs of \$103 million
  - Average PMPY is \$12,264 as shown on the previous chart

## How are the members in the top groups different?

- Randomly sampled 100,000 lives from Marketscan data set for 2005 and 2006
- Sorted the population using three different methods using 2005 as baseline
  - By High Cost Case Model risk score
  - By Prospective All-Encounter DCG-HCC score
  - By 2005 total allowable claims dollars
- Created 1% top-groups for each method (1,000)

# How are the members in the top groups different?

## Top 1% Groups

	% Female	Avg. Age	<u>Hospitalizations</u>		<u>Emergency Room</u>	
			2005	2006	2005	2006
HCCM	55%	50.1	1.1	0.66	1.3	0.93
Prospective DCG	50%	51.3	1.2	0.67	1.2	0.87
Prior Costs	50%	49.5	1.5	0.61	1.5	0.92

	<b>HCCM Prevalence</b>			<b>Prosp. DCG Prevalence</b>			<b>Prior Cost Prevalence</b>		
	2005	2006		2005	2006		2005	2006	
Diabetes	277	240	-13%	357	304	-15%	244	200	-18%
CVD	86	49	-43%	101	66	-35%	94	60	-36%
CHF	122	106	-13%	178	140	-21%	118	92	-22%
COPD	113	80	-29%	146	96	-34%	110	57	-48%
VD	132	88	-33%	160	98	-39%	138	72	-48%
CAD	192	156	-19%	231	171	-26%	251	183	-27%
RF	118	91	-23%	176	121	-31%	80	58	-28%
Respiratory	318	224	-30%	345	222	-36%	296	179	-40%



Aggregated Condition Category Descriptor	HCCM Prevalence	Prosp. DCG Prevalence	Prior Cost Prevalence
ACC001: Infectious and Parasitic	272	270	241
ACC002: Malignant Neoplasm	340	397	297
ACC003: Benign/In Situ/Uncertain Neoplasm	133	116	125
ACC004: Diabetes	277	357	244
ACC005: Nutritional and Metabolic	553	571	564
ACC006: Liver	134	157	115
ACC007: Gastrointestinal	464	463	470
ACC008: Musculoskeletal and Connective Tissue	580	545	590
ACC009: Hematological	354	409	343
ACC010: Cognitive Disorders	54	57	60
ACC011: Substance Abuse	89	100	115
ACC012: Mental	242	205	230
ACC013: Developmental Disability	15	15	24
ACC014: Neurological	248	249	226
ACC015: Cardio-Respiratory Arrest	87	122	117
ACC016: Heart	532	600	601
ACC017: Cerebro-Vascular	86	101	94
ACC018: Vascular	229	262	237
ACC019: Lung	420	452	391
ACC020: Eyes	240	253	230
ACC021: Ears, Nose and Throat	392	365	374
ACC022: Urinary System	365	433	303
ACC023: Genital System	202	194	203
ACC024: Pregnancy Related	12	12	22
ACC025: Skin and Subcutaneous	333	337	316
ACC026: Injury, Poisoning, Complications	433	428	487
ACC027: Symptoms, Signs and Ill-Defined Conditions	802	809	828
ACC028: Neonates	2	9	5
ACC029: Transplants, Openings, Other V-Codes	79	106	77
ACC030: Screening / History	805	799	847



<b>Aggregated RxGroup Category Descriptor</b>	<b>HCCM Prevalence</b>	<b>Prosp. DCG Prevalence</b>	<b>Prior Cost Prevalence</b>
ARXG001: Analgesics/anti-inflammatories	772	699	798
ARXG002: Anti-hyperlipidemics	373	385	359
ARXG003: Anti-infectives	813	785	784
ARXG004: Coagulants and Anticoagulants	237	238	309
ARXG005: Biologicals	186	169	145
ARXG006: Cardiovascular	619	649	621
ARXG007: Neurological agents	679	588	653
ARXG008: Dermatologicals	372	318	310
ARXG009: EENT preparations	275	236	232
ARXG010: Endocrine/metabolic agents	574	480	488
ARXG011: Diabetes drugs	254	227	202
ARXG012: Pulmonary drugs	247	302	220
ARXG013: GI drugs	665	605	596
ARXG014: Genitourinary agents	241	215	205
ARXG015: Immunologic agents	132	107	61
ARXG016: Nutritionals	301	314	254
ARXG017: Upper respiratory agents	397	339	342
ARXG018: Additional groups	324	346	272

## When to use the High Cost Case Model

- When a plan needs to identify the top ½ percent or top 1% of cases expected to be high cost
  - Care management
- When the business problem is:
  - Identifying cases that are going to be catastrophic (high cost) for the plan
    - Pricing, Underwriting
  - Understanding how many and what kinds of stop loss cases are likely to occur (e.g. in a self-insured account)
  - Understanding if there are excess risk coverage or reinsurance considerations



## Recommended Uses of HCCM Top Coding Choices

- “No top coding” – for budgeting and projecting *total* costs
- \$250K and \$100K - when predicting costs below these attachment points
- \$25k - for use by forecasting actuaries and also disease management professionals
  - Model has the best PPV for predicting those likely to exceed \$25k
- HCCM top coding options (250K, 100K and 25K) simulate the impact of reinsurance or stop loss at those levels
  - Top coded models have improved predictive accuracy (as measured by  $R^2$ )

## Applications of high cost claim prediction

- More accurate predictions for individuals & groups
- Group by disease, and then rank
  - DM program involvement
- Rank groups or identify groups with higher concentrations of expected high cost claims
  - Rank by expected year 2 cost
  - Monitoring accounts
- Pooling charges in underwriting or self-insured pricing
- Simulation of reinsurance arrangements or risk pools
  - Better estimate the right tail of the claims distribution

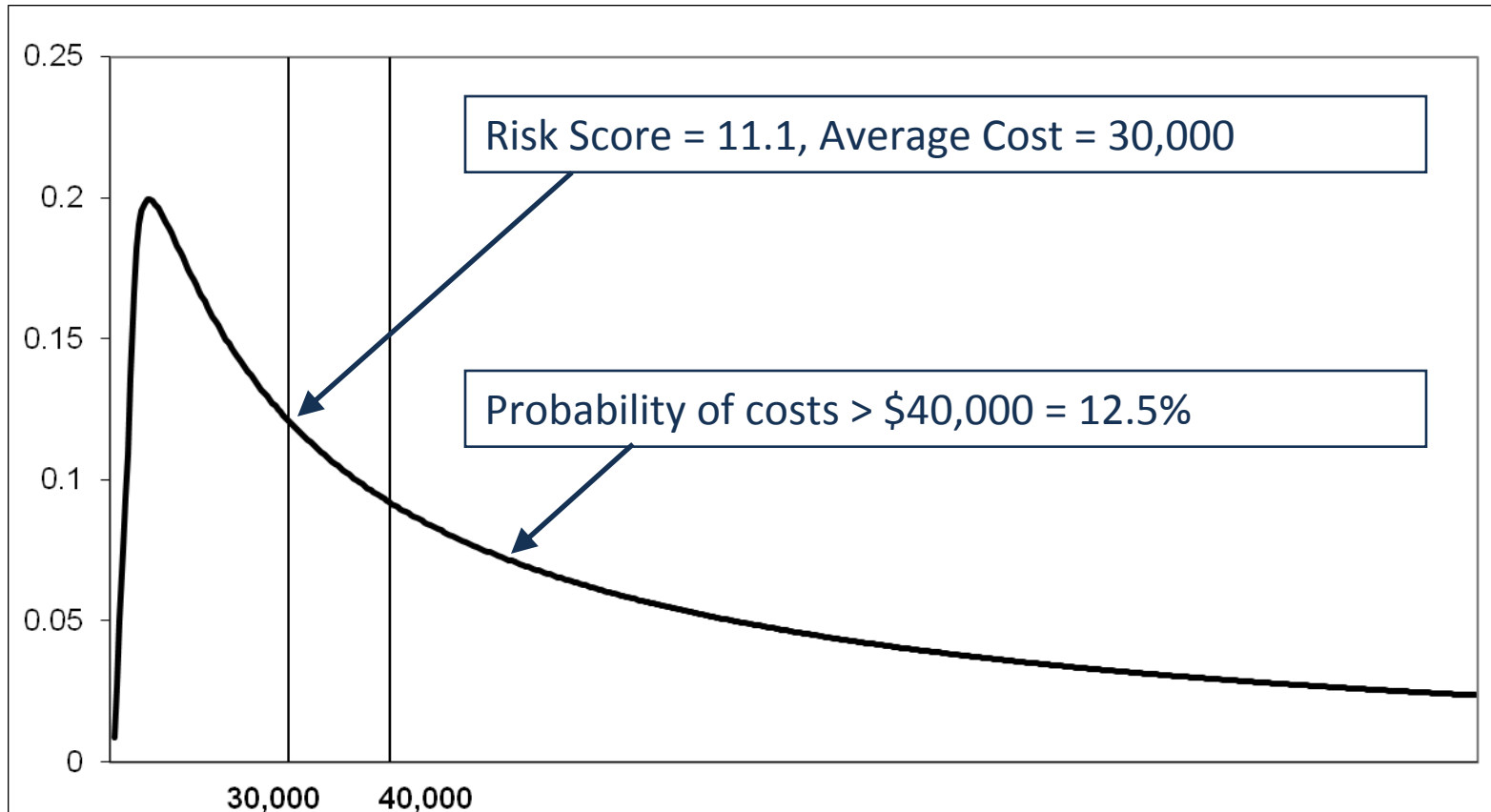
## Reinsurance Considerations

- American Re HealthCare (now Munich Re) gave a user conference presentation in 2004 on high cost claim prediction
  - Evaluated several types of models for predicting high cost claims
    - 2-Part Prospective DCG model with simple recalibration
    - 2-Part Prospective DCG model with “total” recalibration
    - Age-sex tables
    - Prior Costs
    - Claims distributions (e.g., Log-normal, discrete continuance tables)

## Reinsurance Considerations (cont'd)

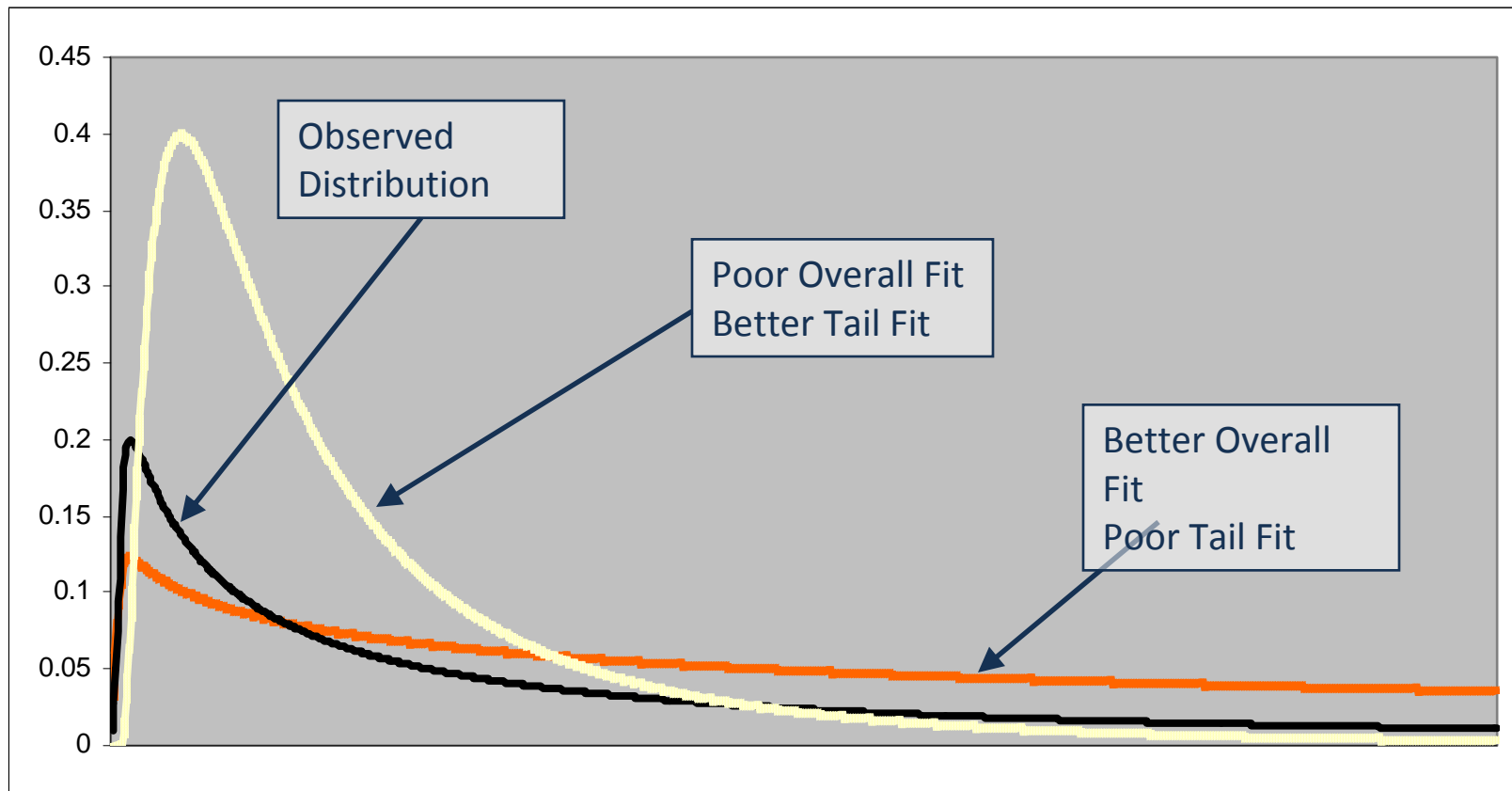
- Risk scores for non-top-coded model reflect total costs
  - You can look at the prevalence of risk scores that would put you over the stop loss threshold (by multiplying by population's average cost)
  - You can look at the prevalence of actual year 2 claims over the stop loss threshold
  - There will be a disconnect!

# Reinsurance Considerations (cont'd)



From American Re “Using DxCG for Stop Loss and Reinsurance Pricing”, 2004 DxCG User Conference Presentation

# Reinsurance Considerations (cont'd)



From American Re “Using DxCG for Stop Loss and Reinsurance Pricing”, 2004 DxCG User Conference Presentation

## American Re retrospective study- methodology

- Methods evaluated:
  - 2-part recalibrations (all HCCs, limited set)
  - Claims distributions based on scores (best fit overall, best fit for top 50%)
  - Age-sex factors
  - Prior year costs
- Looked at ability to identify high cost claimants, excess loss PMPM and grouped R-Squared

## American Re retrospective study- findings

- High cost claim identification
  - Diagnostic models superior in finding high cost claims at all stop loss thresholds
  - Those that the prior cost method successfully identified as high cost had higher excess claims
- PMPM Excess Loss
  - Recalibrated model with limited HCCs was best
  - Prior cost and DxCG raw predictions were equivalent
  - Recalibrated “All HCCs” did not perform well as others



## American Re retrospective study- findings (cont'd)

- Group pricing (PM versus standard methods)
  - Standard methods are age-sex or prior cost
  - Age-sex always worse than diagnostic models
  - Small to mid-size groups (<250): Diagnostic better than prior costs alone (all thresholds)
    - Diagnostic model more limited at \$250K threshold

## American Re retrospective study- findings (cont'd)

- Group pricing (within class of PM)
  - At lower thresholds, recalibrated “All HCCs” better
    - Limited HCCs and distributional models equivalent
  - At \$100K threshold, recalibrate “All HCCs” model and distributional models equivalent
  - At \$250K threshold, the distributional models were better than either of the recalibrated models, though predictive performance was not very strong

## Reinsurance Pooling Scheme

- Large, self-insured employer with national PPO and many Business Units (BUs) each accountable for own healthcare financials
- Corporate decided to “risk-adjust” and bill BUs premiums adjusted to their population
  - Risk premium proxies for Aggregate Stop Loss
  - Billed premiums reconciled with actual claims
  - “Recoveries” paid from Corporate pool, with desired outcome that loss ratios approach 100%



## Pooling Scheme for Self-Insured Employer with Several Business Units

### ***Without Diagnosis-Based Risk Adjustment***

Pooled PPO Claim PMPM	\$250.00			
<b>Business Unit:</b>	<b><u>Corporate</u></b>	<b><u>XYZ Co.</u></b>	<b><u>ABC Co.</u></b>	<b><u>Total</u></b>
Membership	45	455	1,500	2,000
Average Age	48	42	36	38
% Male	70%	25%	70%	60%
Demographic Factor Adjustment	<b>1.15</b>	<b>1.05</b>	<b>0.98</b>	<b>1.00</b>
<i>(Normalized)</i>				
Business Unit Expected PMPM	\$287.50	\$262.50	\$245.08	\$250.00
Risk Pooling Charge as % of Claims	10.0%	7.0%	5.0%	5.6%
Charged Risk Premium PMPM	\$28.75	\$18.38	\$12.25	\$14.02
Total Risk Premium Dollars	\$15,525	\$100,328	\$220,575	<b>\$336,428</b>
Actual Incurred PPO Claims PMPM	\$183.75	\$367.50	\$232.75	\$262.30
Initial Business Unit Loss Ratio	63.9%	140.0%	95.0%	104.9%
Actual minus Expected Claims				
PMPM	(\$103.75)	\$105.00	(\$12.33)	\$12.30
Annual Dollars	(\$56,025)	\$573,300	(\$222,000)	\$295,275
Recoveries collected from Pool	\$0	<b>\$336,428</b>	\$0	\$336,428
Net Owed to the Reinsurance Pool	\$0	\$236,873	(\$222,000)	\$14,873
Final Business Unit Loss Ratio	63.9%	116.5%	95.0%	

## ***Without Diagnosis-Based Risk Adjustment***

### ***Final Pool Accounting***

Starting value	\$0
Risk premium collected	\$336,428
Recoveries paid to units	(\$336,428)
Collected from "losers"	\$236,873
<u>Paid to "winners"</u>	<u>(\$222,000)</u>
<b>Ending value</b>	<b>\$14,873</b>



<b><i>With Diagnosis-Based Risk Adjustment</i></b>						
Pooled PPO Claim PMPM			<b>\$250.00</b>			
Business Unit:			<b><u>Corporate</u></b>	<b><u>XYZ Co.</u></b>	<b><u>ABC Co.</u></b>	<b><u>Total</u></b>
Membership			45	455	1,500	2,000
Average Age			48	42	36	38
% Male			70%	25%	70%	60%
Demographic Factor Adjustment			1.15	1.05	0.98	1.00
		<i>(Normalized)</i>				
"Hidden Health Information"			<i>Low incidence of chronic disease for this age group</i>	<i>High incidence of diabetes, heart disease and associated comorbidities</i>	<i>Young, healthy and invincible</i>	
Unit's Average Relative Risk Score			0.85	1.40	0.88	1.00
		<i>(Prospective, Normalized)</i>				



<b><i>With Diagnosis-Based Risk Adjustment</i></b>					
Pooled PPO Claim PMPM		<b>\$250.00</b>			
Business Unit:		<b><u>Corporate</u></b>	<b><u>XYZ Co.</u></b>	<b><u>ABC Co.</u></b>	<b><u>Total</u></b>
Membership		45	455	1,500	2,000
Average Age		48	42	36	38
% Male		70%	25%	70%	60%
Unit's Average Relative Risk Score		0.85	1.40	0.88	1.00
	<i>(Prospective, Normalized)</i>				
Business Unit Risk-Adjusted PMPM		\$212.50	\$350.00	\$220.79	\$250.00
Risk Pooling Charge as % of Claims		10.0%	7.0%	5.0%	5.7%
	<i>Charged Risk Premium PMPM</i>	\$21.25	\$24.50	\$11.04	\$14.33
	<i>Total Risk Premium Dollars</i>	\$11,475	\$133,770	\$198,713	<b>\$343,958</b>
Actual Incurred PPO Claims PMPM		\$183.75	\$367.50	\$232.75	\$262.30
Initial Business Unit Loss Ratio		86.5%	105.0%	105.4%	104.9%
Actual minus Expected Claims					
	PMPM	<b>(\$28.75)</b>	\$17.50	\$11.96	\$12.30
	Annual Dollars	<b>(\$15,525)</b>	\$95,550	\$215,250	\$295,275
Recoveries collected from Pool		\$0	<b>\$95,550</b>	<b>\$215,250</b>	\$310,800
Net Owed to the Reinsurance Pool		\$0	\$0	\$0	\$0
Final Business Unit Loss Ratio		86.5%	100.0%	100.0%	

## ***With Diagnosis-Based Risk Adjustment***

### ***Final Pool Accounting***

Starting value	\$0
Risk premium collected	\$343,958
Recoveries paid to units	(\$310,800)
Collected from "losers"	\$0
<u>Paid to "winners"</u>	<u>\$0</u>
<b>Ending value</b>	<b>\$33,157</b>





# Any Questions?

Vincent Kane, FSA, MAAA  
Research Scientist  
DxCG – A Division of Urx, Inc.  
[vincent.kane@dxcg.com](mailto:vincent.kane@dxcg.com)  
[www.dxcg.com](http://www.dxcg.com)