Data Preparation

Analysis of combined_data.csv

Sample Selection

Item	Amount
# of Samples	4619
# of Samples with Purchases	1411

Attribute Creation

A new categorical attribute was created to enable analysis of players as broken into 2 categories (HighRollers and PennyPinchers). A screenshot of the attributes follows:

File

Row ID	userId	userSe	teamLevel	S platfor	count	count	count	D avg_price	S avg_pri
Row4	937	5652	1	android	39	0	1	1	PennyPincher
Row11	1623	5659	1	iphone	129	9	1	10	HighRollers
Row13	83	5661	1	android	102	14	1	5	PennyPincher
Row17	121	5665	1	android	39	4	1	3	PennyPincher
Row18	462	5666	1	android	90	10	1	3	PennyPincher
Row31	819	5679	1	iphone	51	8	1	20	HighRollers
Row49	2199	5697	1	android	51	6	2	2.5	PennyPincher
Row50	1143	5698	1	android	47	5	2	2	PennyPincher
Row58	1652	5706	1	android	46	7	1	1	PennyPincher
Row61	2222	5709	1	iphone	41	6	1	20	HighRollers
Row68	374	5716	1	android	47	7	1	3	PennyPincher
Row72	1535	5720	1	iphone	76	7	1	20	-
VALUE AND DESCRIPTION OF THE PERSON OF THE P		0022206.0		************	5,000	Total State of the Control of the Co	266	2000	HighRollers
Row73	21	5721	1	android	52	2	1	3	PennyPincher
Row101	2379	5749	1	android	62	9	1	3	PennyPincher
Row122	1807	5770	1	iphone	177	25	2	7.5	HighRollers
Row127	868	5775	1	iphone	54	5	1	10	HighRollers
Row129	1567	5777	1	android	27	4	2	4	PennyPincher
Row131	221	5779	1	iphone	37	2	1	20	HighRollers
Row135	2306	5783	1	android	67	5	1	1	PennyPincher
Row137	1065	5785	1	iphone	37	5	2	11.5	HighRollers
Row140	827	5788	1	iphone	75	5	1	20	HighRollers
Row150	1304	5798	1	mac	71	9	2	11.5	HighRollers
Row158	1264	5806	1	linux	81	12	1	5	PennyPincher
Row159	1026	5807	1	iphone	52	10	1	20	HighRollers
Row163	649	5811	1	linux	51	9	1	1	PennyPincher
Row169	1958	5817	1	android	40	3	1	20	HighRollers
Row172	1300	5820	1	android	58	1	2	3	PennyPincher
Row 186	178	5834	1000	CONTRACTOR CONTRACTOR	54	6	1	20	HighRollers
Market Control	200000	C (C	1	iphone	200	N (*) - (.)	333	200	
Row196	670	5844	1	iphone	38	3	2	20	HighRollers
Row207	208	5855	1	iphone	32	3	1	20	HighRollers
Row210	157	5858	1	iphone	32	2	1	10	HighRollers
Row212	2221	5860	1	iphone	191	18	2	11.5	HighRollers
Row215	471	5863	1	iphone	45	6	2	15	HighRollers
Row218	1234	5866	1	android	46	3	1	10	HighRollers
Row222	371	5870	1	android	53	9	1	3	PennyPincher
Row232	2146	5880	1	linux	46	7	1	2	PennyPincher
Row239	935	5887	1	iphone	57	2	1	10	HighRollers
Row241	165	5889	1	iphone	49	3	1	5	PennyPincher
Row244	1538	5892	1	iphone	24	3	1	20	HighRollers
Row245	1544	5893	1	iphone	36	6	2	20	HighRollers
Row261	2218	5909	1	android	80	6	1	3	PennyPincher
Row262	1162	5910	1	windows	192	16	1	2	PennyPincher
Row266	1821	5914	1	windows	178	22	1	1	PennyPincher
Row200	2133	5919	1	android	87	14	1	3	PennyPincher
		100000			530	1 - 1	1		
Row272	1027	5920	1	iphone	52	5	3	15	HighRollers
Row273	518	5921	1	linux	121	16	1	1	PennyPinche
Row282	2029	5930	1	iphone	89	7	1	10	HighRollers
Row286	2384	5934	1	windows	41	5	1	1	PennyPincher
Row290	1155	5938	1	iphone	71	16	1	20	HighRollers
Row292	564	5940	1	linux	34	3	1	1	PennyPincher
Row293	97	5941	1	android	74	10	1	1	PennyPinche
Row297	253	5945	1	android	66	12	1	3	PennyPinche
Row302	934	5950	1	windows	43	4	1	10	HighRollers
Row307	2009	5955	1	iphone	27	4	1	5	PennyPincher
Row324	1815	6088	1	iphone	272	24	1	3	PennyPincher
Row325	864	6173	1	android	79	9	1	3	PennyPinche

The avg_price_binned attribute was created by binning on the avg_price attribute and distinguishes between "HighRollers" and "PennyPinchers." HighRollers are buyers of items that cost more than \$5.00. PennyPinchers are buyers of items that cost \$5.00 or less.

The creation of this new categorical attribute was necessary because we want to separate players into two categories based on their purchasing behavior.

Attribute Selection

The following attributes were filtered from the dataset for the following reasons:

Attribute	Rationale for Filtering
platformType	To identify the platform most used by HighRollers and PennyPinchers
count_hits	To identify the number of game hits achieved by HighRollers and PennyPinchers
avg_price	To identify HighRollers and PennyPinchers

Data Partitioning and Modeling

The data was partitioned into train and test datasets.

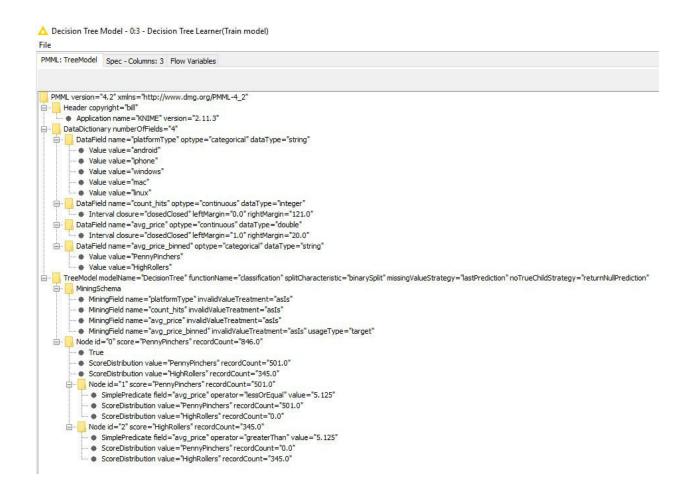
The train data set was used to create the decision tree model.

The trained model was then applied to the test dataset.

This is important because we want to test our model on data that was not used to train (i.e., create) it.

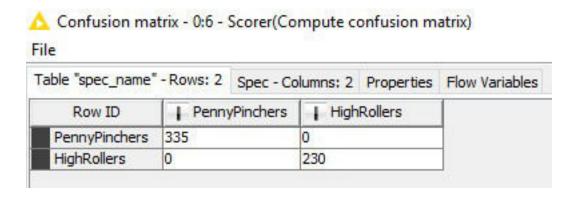
When partitioning the data using sampling, it is important to set the random seed because we want to get reproducible results.

A screenshot of the resulting decision tree can be seen below:



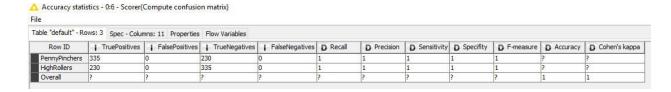
Evaluation

A screenshot of the confusion matrix can be seen below:



As seen in the screenshot above, the overall accuracy of the model is 100%

A screenshot of the accuracy statistics can be seen below:



There are 335 PennyPinchers and 0 HighRollers for the PennyPinchers category. There are 230 HighRollers and 0 PennyPinchers for the HighRollers category.

Analysis Conclusions The final KNIME workflow is shown below:

This Workflow uses a File Reader node to import the combined data dataset. The Row Filter node then filters the dataset to retain only rows with purchases and identify that number. The Numeric Binner node creates an avg_price_binned attribute based on the values in the avg_price attribute (HighRollers have an avg_price value over 5 and PennyPinchers have an avg_price of 5 or less). The Column Filter node filters the dataset to retain only the relevant columns (platformType, count_hits, avg_price and avg_price_binned). The Color Manager assigns the color red to the PennyPinchers and blue to the HighRollers. The Partitioning node partitions the dataset into train and test datasets (60% train and 40% test), using Stratified sampling with a random seed value of 1466016757670. The Decision Tree Learner node accepts and processes the train dataset. The Decision Tree Predictor node accepts and processes the train and test datasets. The Scorer node accepts that output and generates the confusion matrix and accuracy statistics. Decision Tree Learner 地上 0.00 000 0.00 000 0.00 Categorize HighRollers Filter for Assign colors Train model ecision Tree and PennyPinchers relevant File Reader Predictor Scorer 0.00 Split data 60/40 (m.m.) Apply model Compute confusion matrix

What makes a HighRoller vs. a PennyPincher?

We defined a HighRoller as a user who buys items that cost more than \$5.00 and a PennyPincher as a user who buys items that cost \$5.00 or less.

Building on our observations from the Week1Technical Appendix, where we found the top 3 spenders use the iPhone platform, we can now confirm that the majority of HighRollers use the iPhone platform according to the Classified Data found in the Decision Tree Predictor. We did not find any reasonable correlation between count_hits and HighRollers or PennyPinchers in this analysis.

A screenshot of the Classified Data can be seen below.

Row ID	S platfor	count	D avg_price	S avg_pri	S Predicti
Row17	android	4	3	PennyPinchers	PennyPincher
Row58	android	7	1	PennyPinchers	PennyPincher
Row61	iphone	6	20	HighRollers	HighRollers
Row73	android	2	3	PennyPinchers	_
Row101	android	9	3	PennyPinchers	
Row127	iphone	5	10	HighRollers	HighRollers
Row129	android	4	4	PennyPinchers	
Row135	android	5	1	PennyPinchers	
Row150	mac	9	11.5	HighRollers	HighRollers
Row 158	linux	12	5	PennyPinchers	
Row159	iphone	10	20	HighRollers	HighRollers
Row 172	android	1	3	PennyPinchers	
Row207	iphone	3	20	HighRollers	HighRollers
Row212	iphone	18	11.5	HighRollers	HighRollers
Row212		7	2	PennyPinchers	-
Row232 Row241	linux	3	5	PennyPinchers PennyPinchers	
NAME OF TAXABLE PARTY.	iphone		9.5		
Row245	iphone	6	20	HighRollers	HighRollers
Row262	windows	16	2	PennyPinchers	
Row282	iphone	7	10	HighRollers	HighRollers
Row290	iphone	16	20	HighRollers	HighRollers
Row293	android	10	1	PennyPinchers	
Row325	android	9	3	PennyPinchers	
Row328	windows	4	1	PennyPinchers	
Row344	windows	1	3	PennyPinchers	
Row349	iphone	1	10	HighRollers	HighRollers
Row358	iphone	16	20	HighRollers	HighRollers
Row368	android	23	2	PennyPinchers	PennyPincher
Row369	windows	5	5	PennyPinchers	PennyPincher
Row371	android	13	3	PennyPinchers	PennyPincher
Row380	iphone	5	6	HighRollers	HighRollers
Row386	iphone	11	7.5	HighRollers	HighRollers
Row387	iphone	6	3	PennyPinchers	PennyPincher
Row391	linux	7	1	PennyPinchers	PennyPincher
Row392	iphone	2	12.5	HighRollers	HighRollers
Row395	iphone	22	15	HighRollers	HighRollers
Row408	android	3	20	HighRollers	HighRollers
Row440	android	5	11.5	HighRollers	HighRollers
Row441	windows	5	2	PennyPinchers	The state of the s
Row450	linux	8	2	PennyPinchers	
Row455	iphone	7	5	PennyPinchers	
Row466	iphone	3	20	HighRollers	HighRollers
Row482	android	11	3	PennyPinchers	-
Row530	android	9	6.5	HighRollers	HighRollers
Row533	android	4	3	PennyPinchers	The state of the s
Row541	windows	10	1	PennyPinchers	
Row553	iphone	8	20	HighRollers	HighRollers
ACCURATION AND ADDRESS.	android	20	3		-
Row561	20 00 00 00 00 00 00 00 00 00 00 00 00 0	11	445	PennyPinchers	
Row582	iphone	2	20	HighRollers	HighRollers
Row589	windows	28	10	HighRollers	HighRollers
Row605	linux	18	1	PennyPinchers	
Row616	windows	12	1	PennyPinchers	
Row636	windows	30	1	PennyPinchers	
Row641	linux	11	1	PennyPinchers	
Row646	mac	4	10	HighRollers	HighRollers
Row650	android	14	3	PennyPinchers	The second secon
Row653	windows	7	1	PennyPinchers	PennyPincher

Specific Recommendations to Increase Revenue

- 1. Given that the HighRollers typically use iPhones, it would be beneficial to offer more ads for iPhone-related products in general.
- 2. Assuming it is possible to generate targeted ads, offering ads for in-game purchases (e.g., binoculars) to players with low count_hits should increase revenue.