

Prediction Markets: Do They Improve Risk Management?

Jon Schuler

MITRE Corporation

jschuler@mitre.org

703-983-3884

22 September 2010

What we did:

- Introduced an *Inklings*TM prediction market to the USAF
- Generic questions:
 - “Will the New England Patriots sign Terrell Owens by the start of the 2010 NFL season? “
 - “Will both BP and U.S. Government sources officially report no more oil is leaking from the Deepwater Horizon well by July 1, 2010?”
- Acquisition-program specific questions:
 - “Will Predator platforms adopt the Program2 Rev B specs?”
 - “Will any of the load-balancing, multi-path TRILL boxes tested at last month's UNH Interop lab appear in an USAF Advanced Technology Demonstration by Spring 2011?”
 - “Which of following Spectrum bands will Program2 be approved for use by Jan 2011”

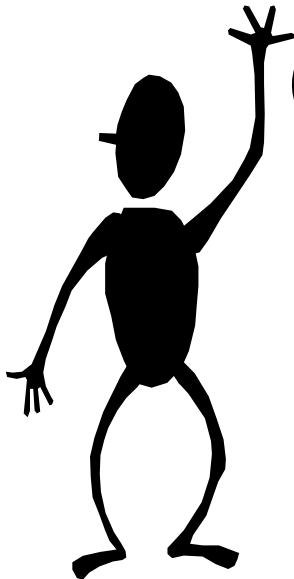
“What’s a prediction market good for?”

Improve dialogue within
acquisition team

Lets people offer
“reprisal-free” feedback

Foster diversity of viewpoints
into analysis

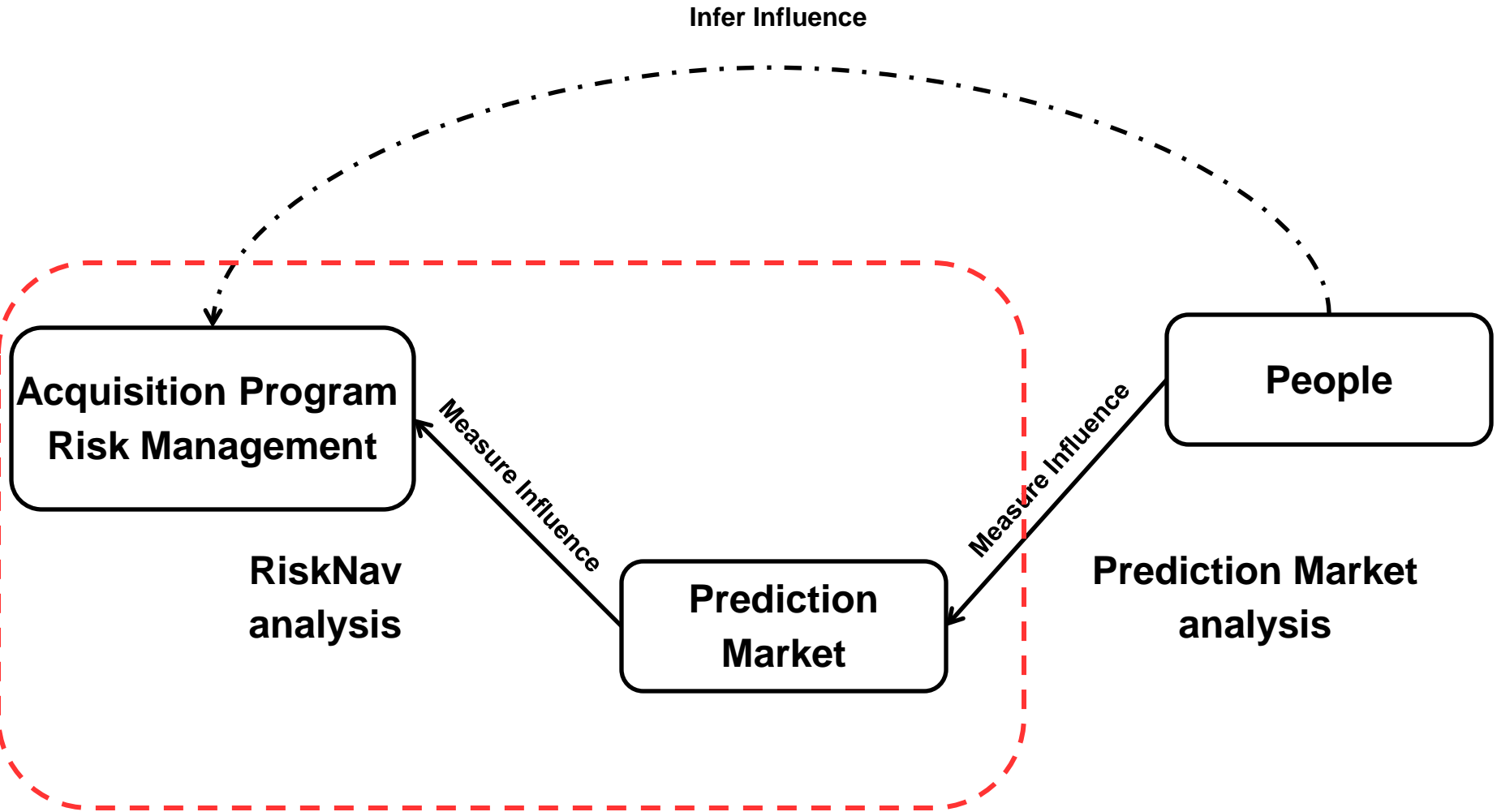
Etc. etc etc.



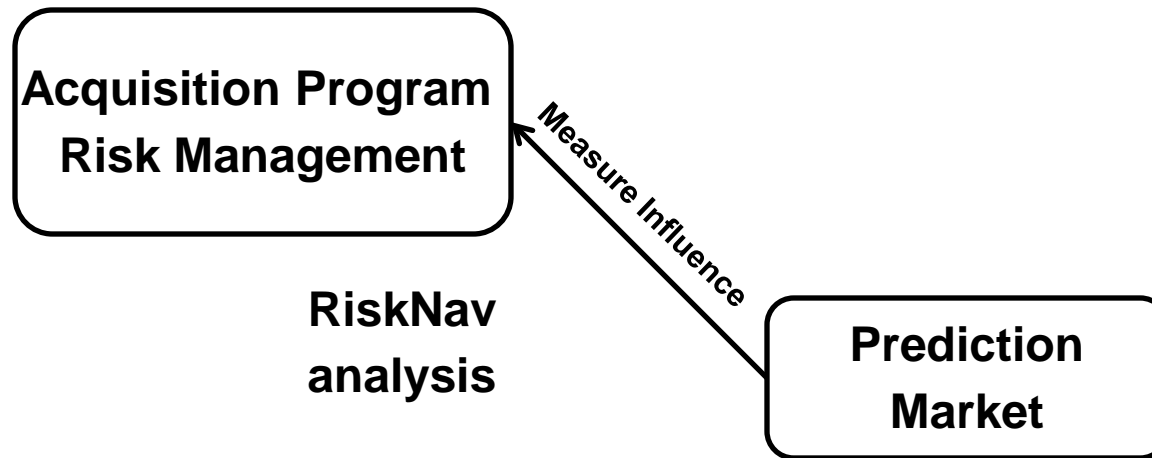
Can we clinically measure that?

- **Pre-existing risk-management process already in place**
 - MITRE provides on-site support to the USAF
 - Established process that identifies, enunciates, quantifies, and mitigates acquisition risks
 - 14 acquisition programs; various levels of activity
- **2+ years data logged in RiskNav software**
 - Front-end: web-based interface
 - Enter/modify assessments. Provide summary display
 - Back-end: Microsoft Access database
 - Primary use: characterize current state
 - Mining DB change-logs can extract a historical record

Prediction Market Analysis Framework

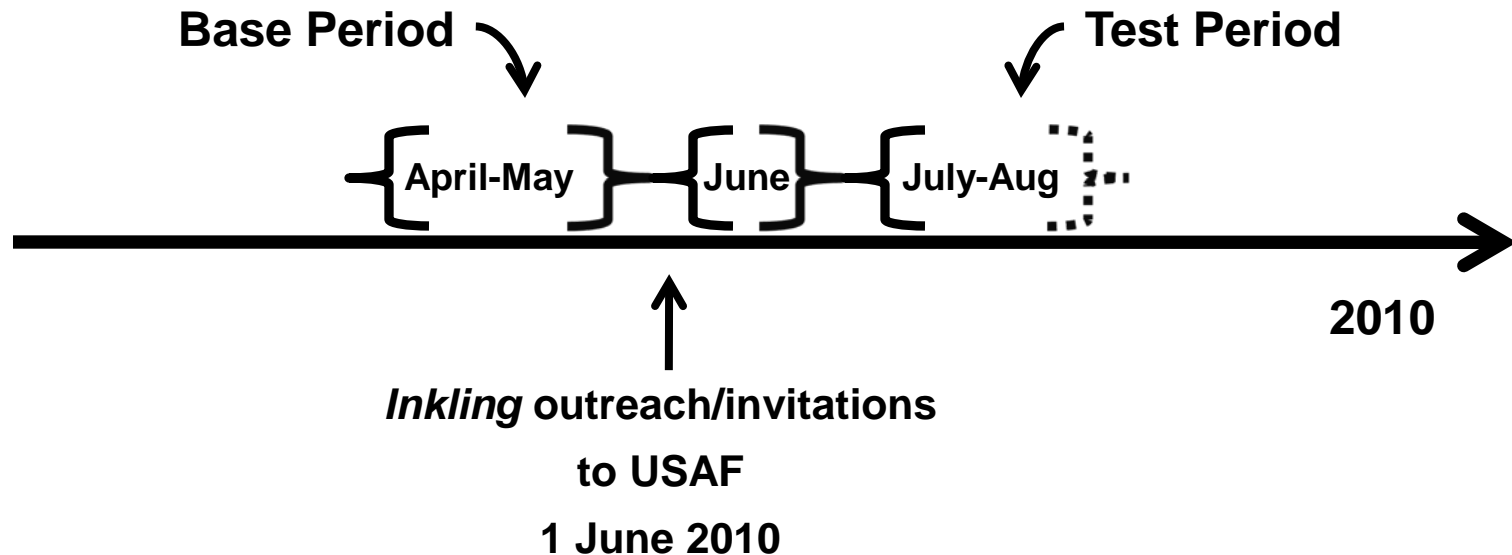


'Treatment Effects' we would expect to see



- Is there an increase in the overall database activity?
- Is there an increase in the rate of newly identified risks?
- Are new risks identified earlier from their event-horizon?
- Do risks get mitigated or closed more quickly?

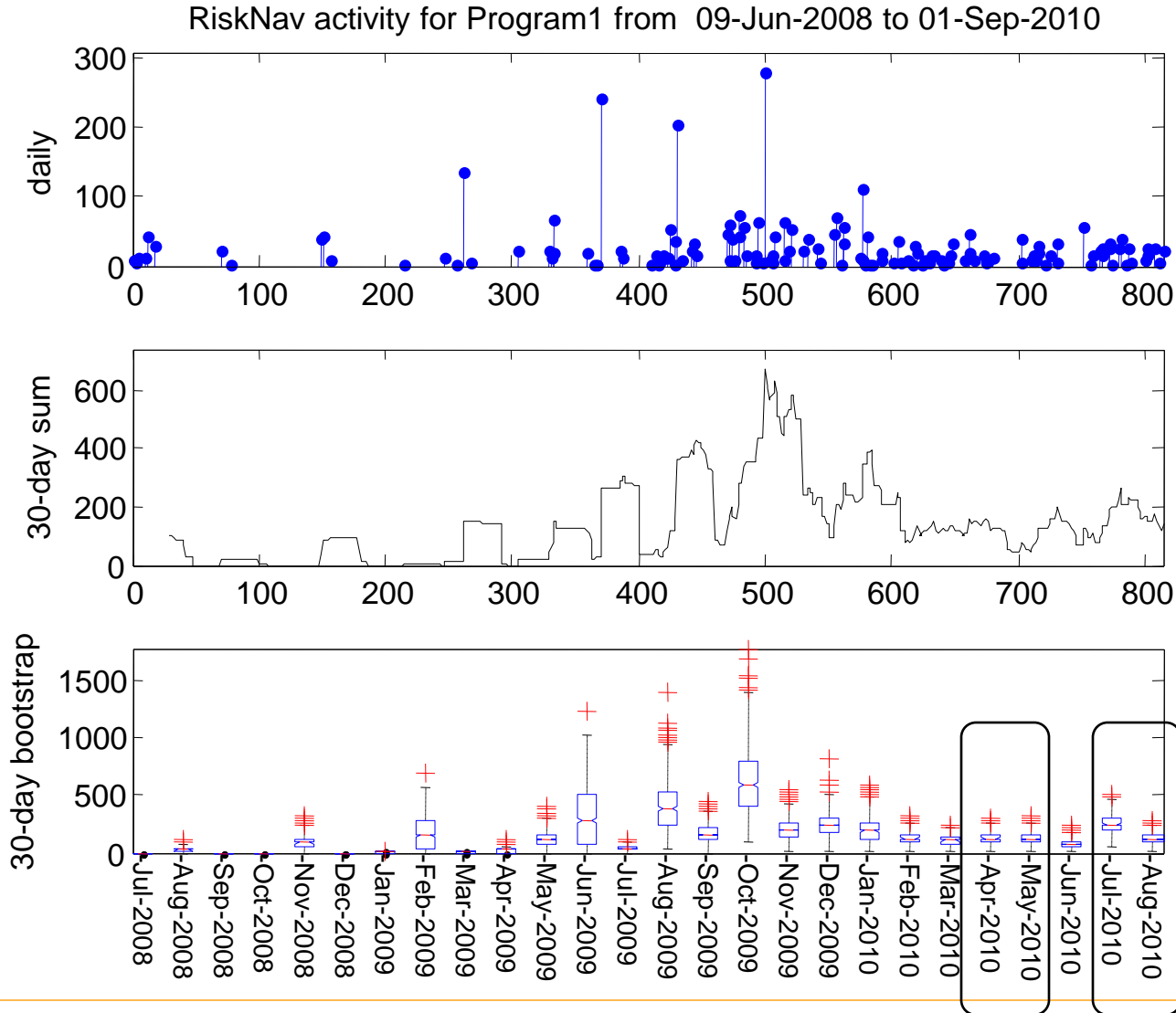
Event timeline under analysis



■ Special Note:

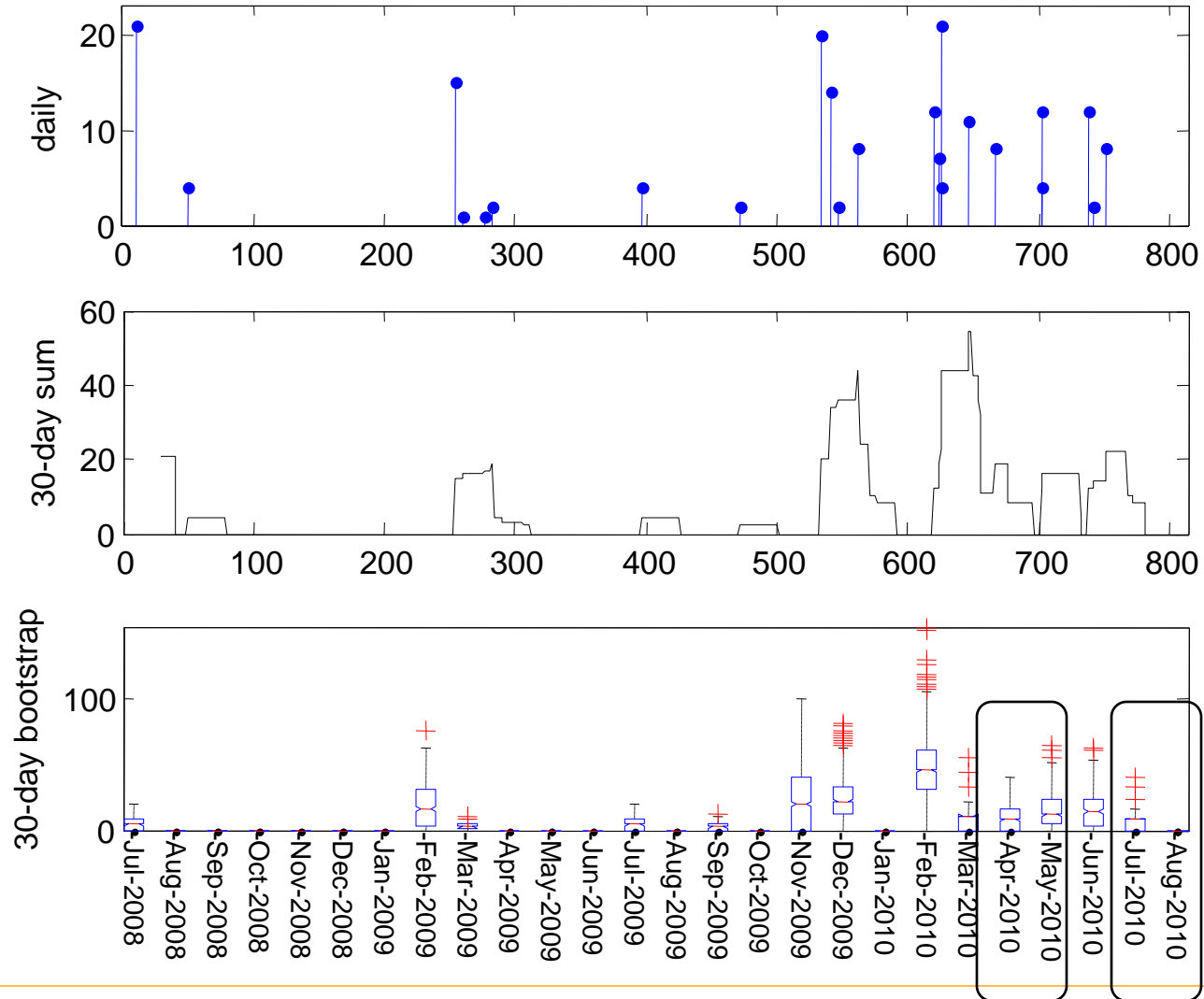
- This doesn't analyze the prediction market itself; we just assume that one took place beginning June 2010
- This analyzes project risk-management activity, comparing 2 months before with 2 months after the prediction market began

Database activity over time



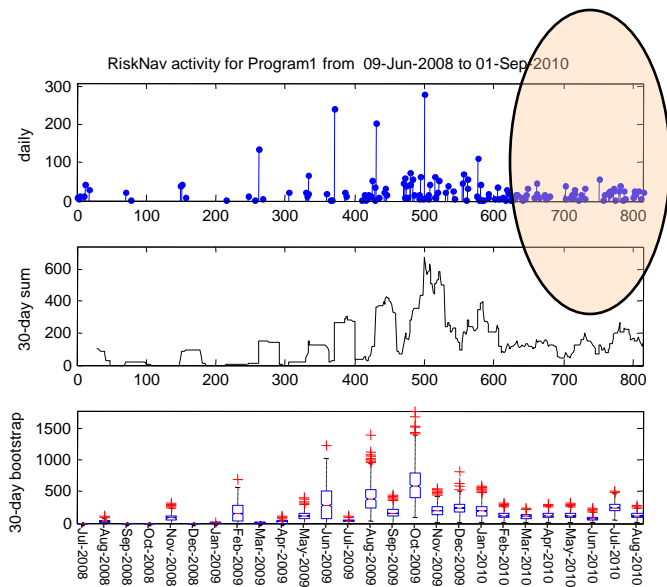
Database activity over time

RiskNav activity for Program5 from 09-Jun-2008 to 01-Sep-2010

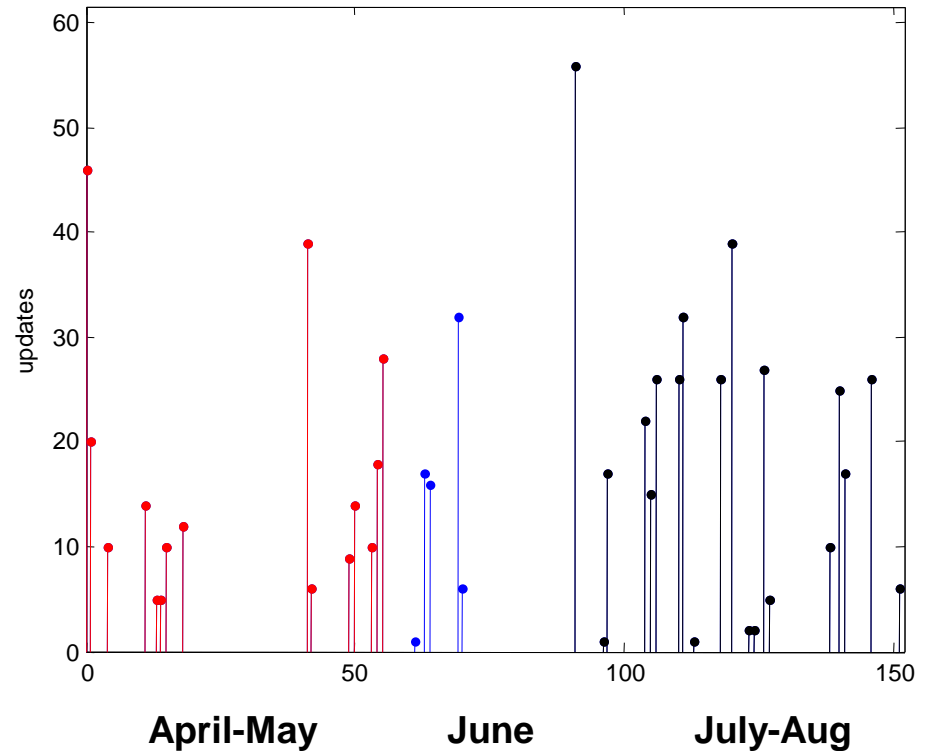


Program1 update activity

- April-May updates: 246 (Baseline)
- July-Aug updates: 381 (Test period)



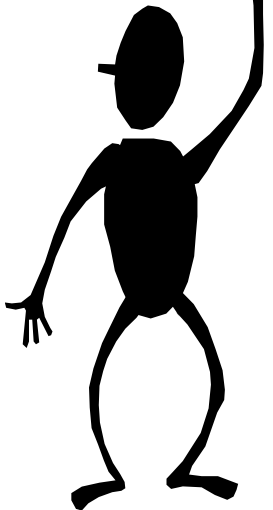
RiskNav activity for Program1 from 01-Apr-2010 to 31-Aug-2010



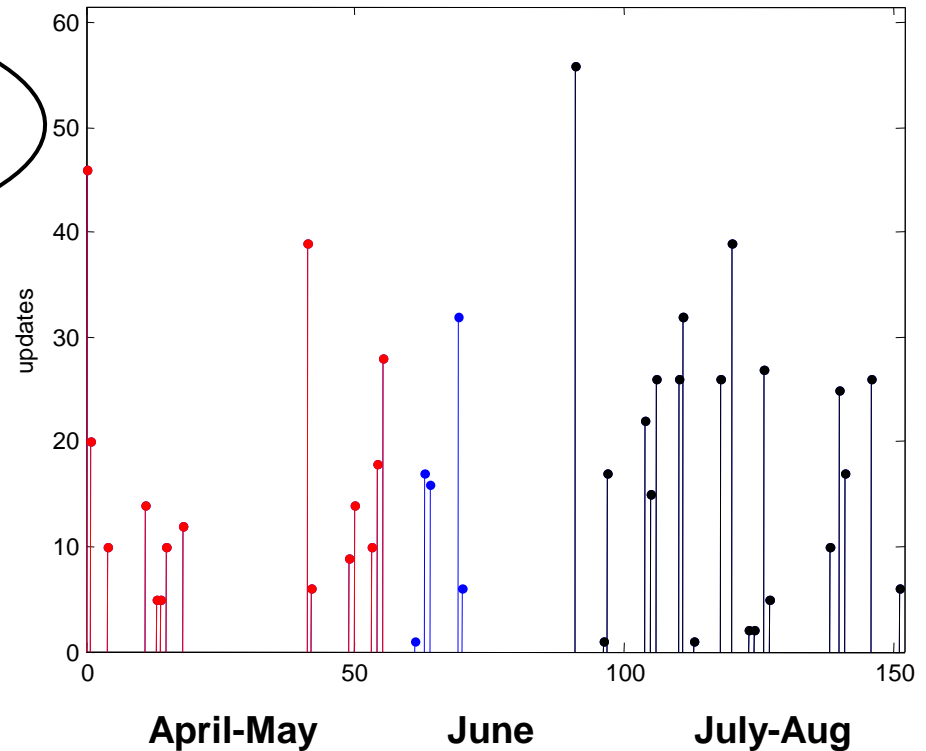
Program1 update activity

- April-May updates: 246 (Baseline)
- July-Aug updates: 381 (Test period)

O.K....I can see an increase in activity but is that increase *significant*?

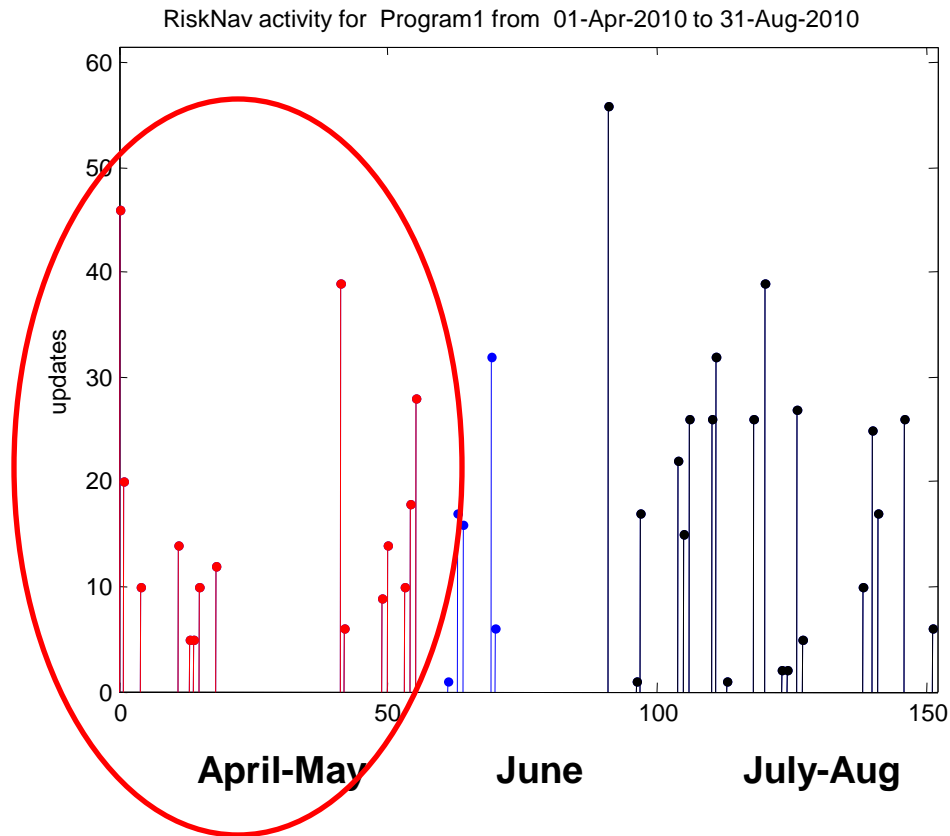


RiskNav activity for Program1 from 01-Apr-2010 to 31-Aug-2010



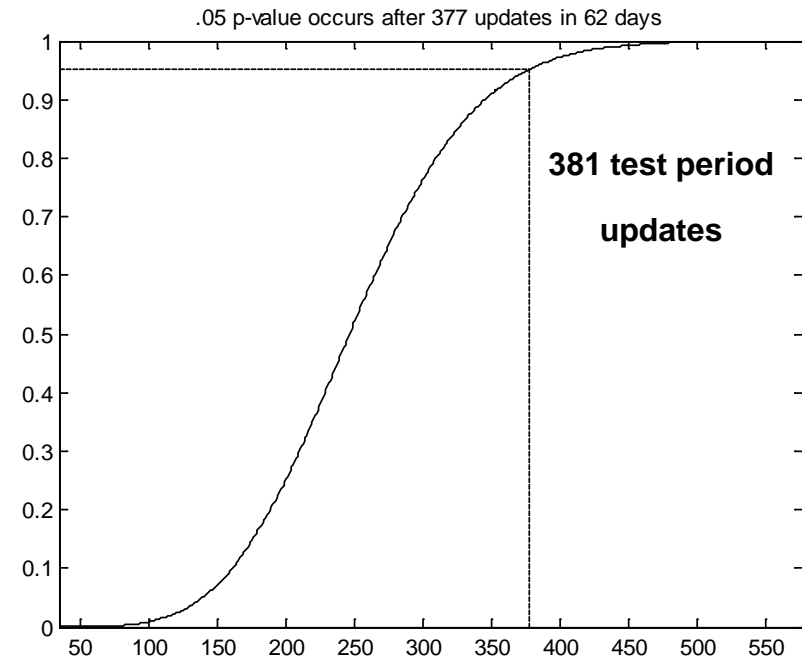
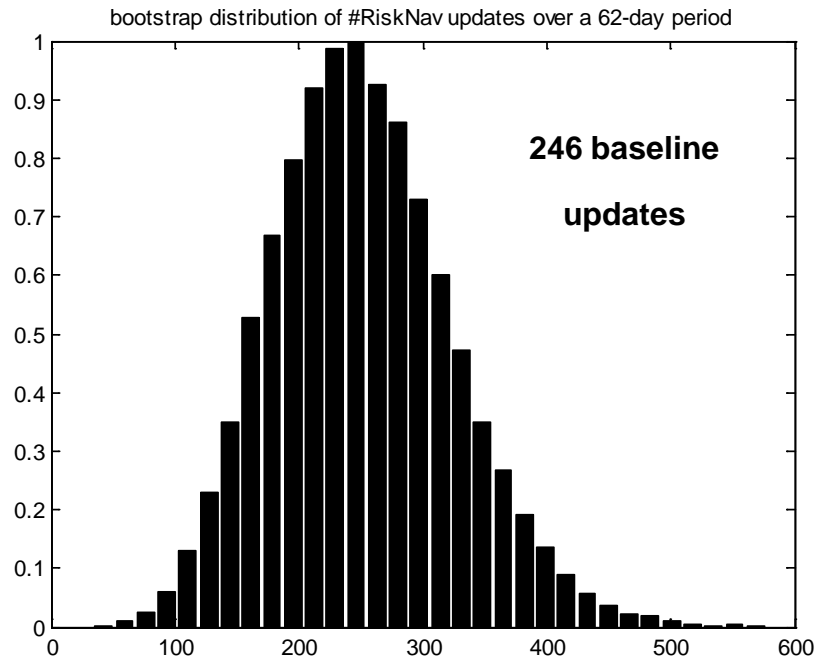
Bootstrap re-sampling of baseline (Apr-May) data

- Pick 62 days at random (with replacement) from baseline period
 - Tally up the total number of database updates observed in re-sample



Bootstrap re-sampling of baseline (Apr-May) data

- Pick 62 days at random (with replacement) from baseline period
 - Tally up the total number of database updates observed in re-sample
- Repeat a few thousand times; build empirical distribution

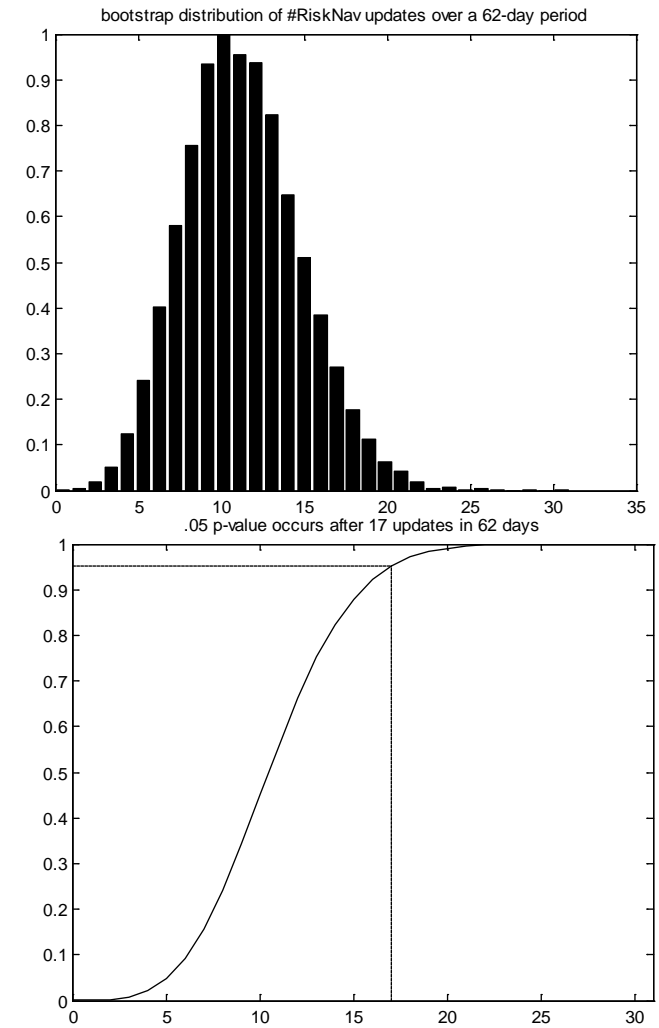
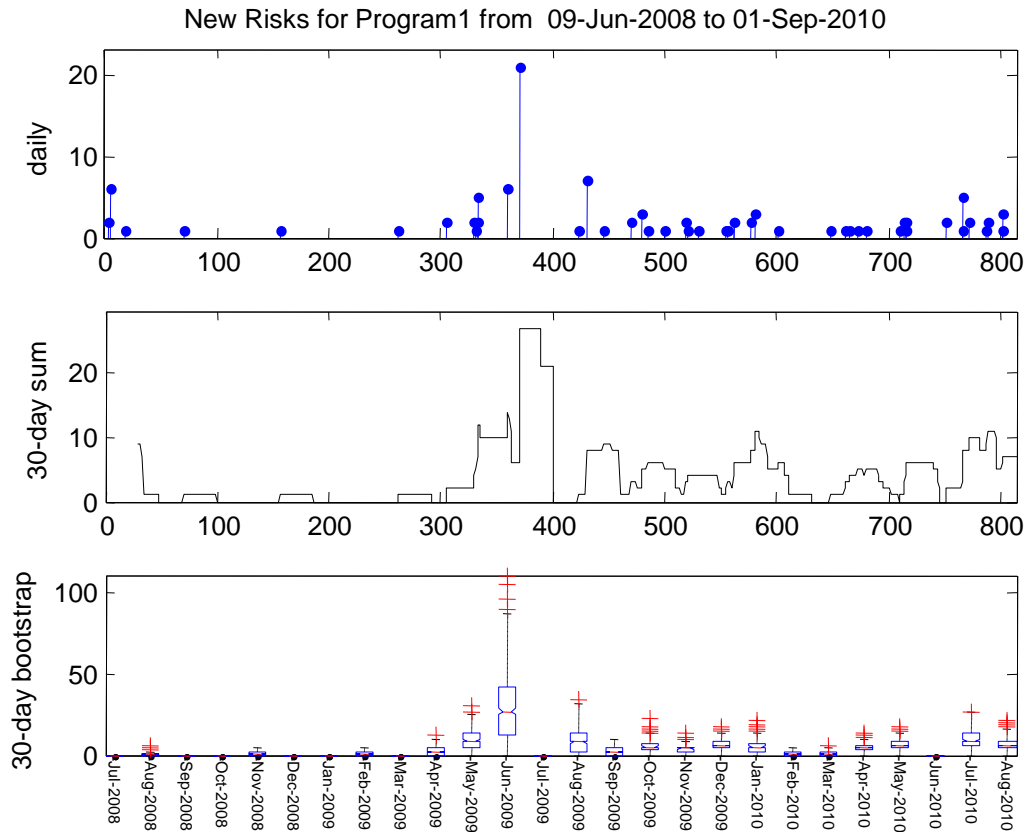


■ 1-sided hypothesis test

- If we see 377 or more RiskNav updates in 62 days, we can ascribe this increase to the Prediction Market...at the 5% chance this could have otherwise occurred 'naturally'

Examine the rate of newly identified risks

- Corresponding bootstrap analysis: how many new risks must we see in a 62-day period?

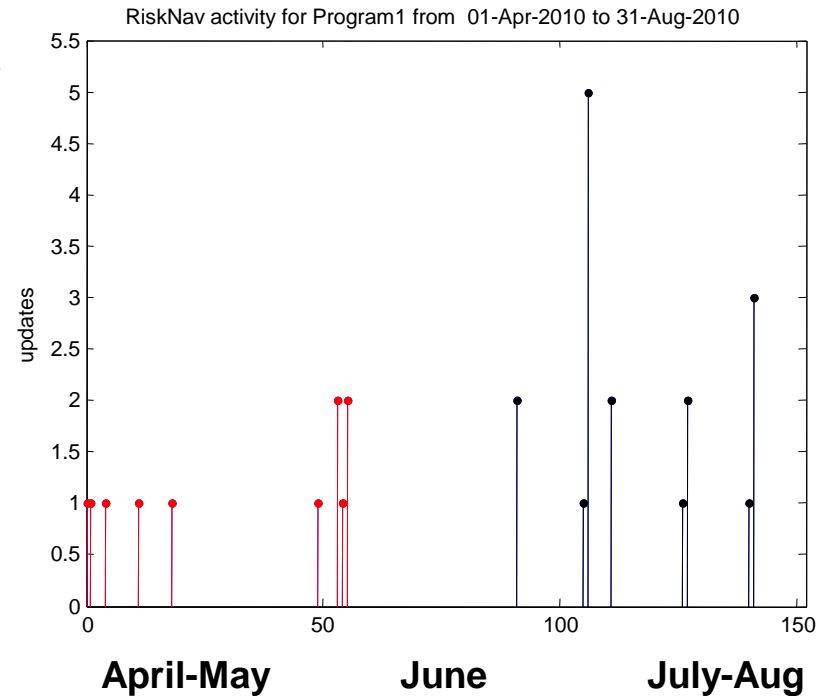
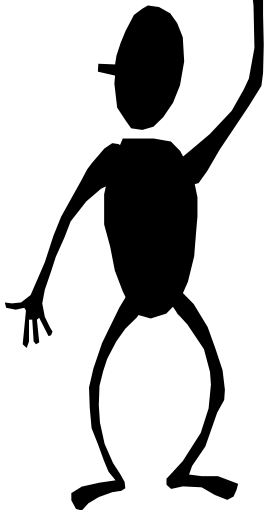


- Concluding answer: need to see 17+ new risks

Program1 new risks

- April-May new risks: 11 (Baseline)
- July-Aug new risks: 17 (Test period)

O.K....I can see an increase in activity,
and it is statistically anomalous

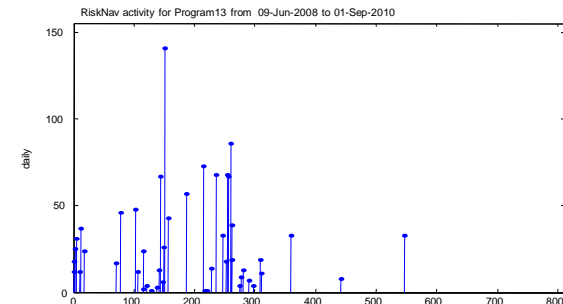
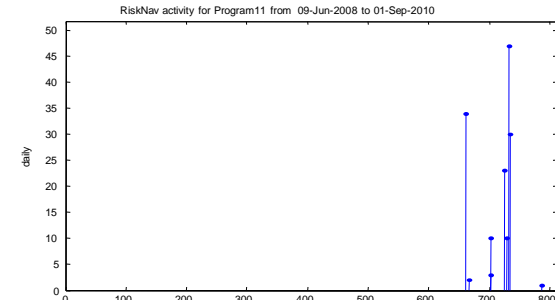
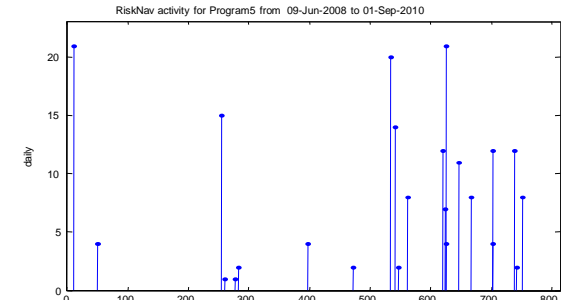
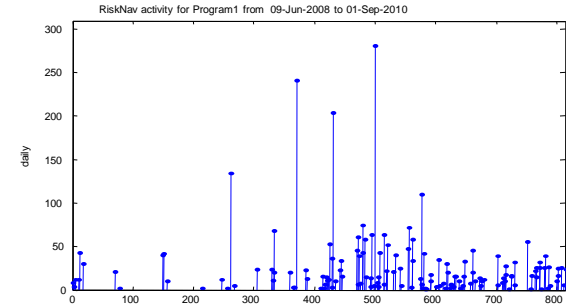


Update & New Risk activity for all programs

Risk Management		Update	New	# Observed Updates				# Observed New Risks			
Process Maturity	Program	Activity	Risks	Baseline	Test	.05 Level	p-value	Baseline	Test	.05 Level	p-value
Self-Generating	Program1	Y	Y	246	381	376	0.044	11	17	17	0.047
	Program2	*	Y	258	117	377	0.991	6	8	6	0.003
	Program3	**		162	145	246	0.635	3	0	3	0.640
Stimulus Driven	Program4	Y	Y	100	222	173	0.003	0	5	0	0.000
	Program5			24	8	52	0.839	1	0	3	0.646
	Program6	Y		8	21	20	0.027	3	0	0	0.000
	Program7			63	56	132	0.541	2	1	7	0.738
	Program8	Y	Y	10	30	24	0.012	0	2	0	0.000
Initiating	Program9			0	0	0	0.000	0	0	0	
	Program10		Y	0	20	0	0.000	0	3	0	0.000
	Program11			46	1	117	0.967	3	0	7	0.870
Inactive	Program12			0	0	0		0	0	0	
	Program13			0	0	0		0	0	0	
	Program14			0	0	0		0	0	0	

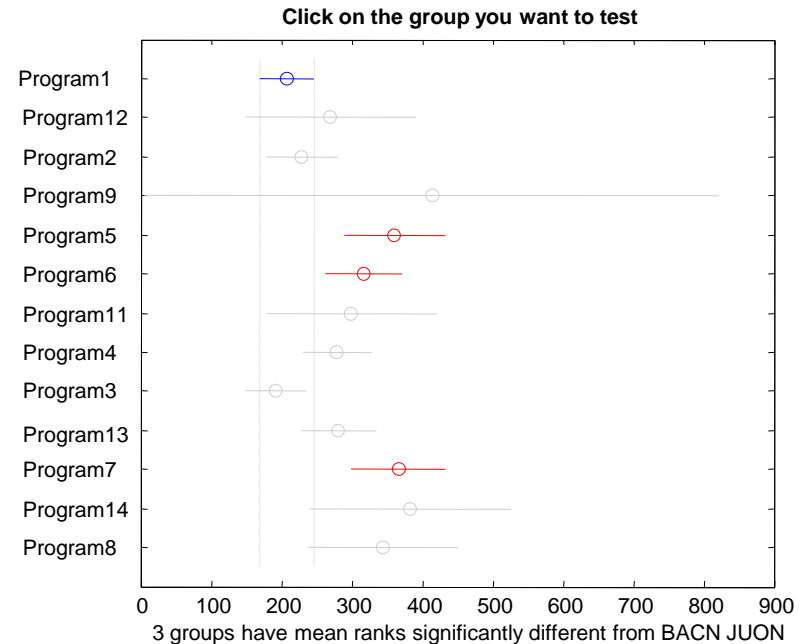
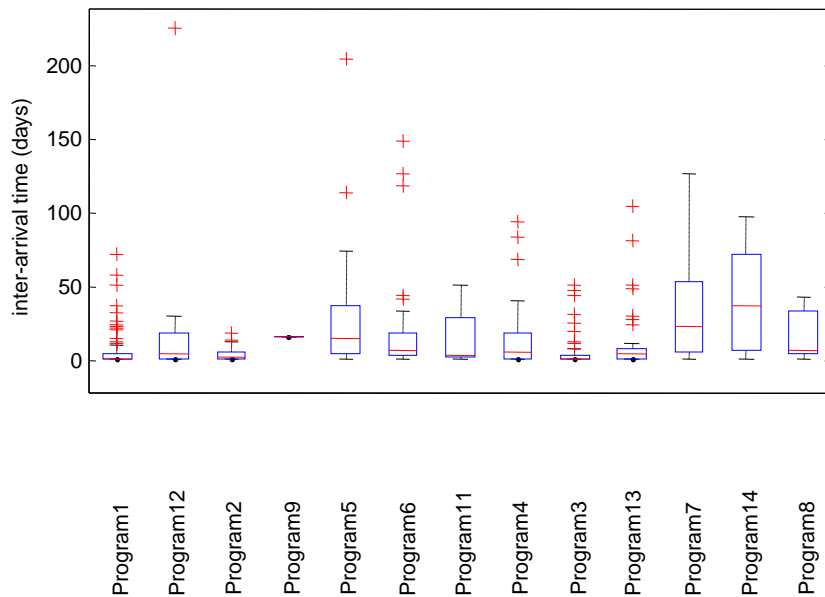
Let's break this down into something simpler

- 4 types of risk-management cultures observed in the 14 programs
 - **Self-generating** - routine updates on a consistent basis with no evident stimulus
 - **Stimulus Driven** – periodic updates as the result of external events (e.g. PMRs, risk meetings, risk team interactions)
 - **Initiating** – programs initializing a risk management process or use of the tool
 - **Inactive** – programs no longer actively managing risks



You can distinguish programs rigorously...

- Look at the #days between database entries
- Kruskal-Wallis rank-based 1-way ANOVA tells demarks programs by significantly different median inter-arrival times



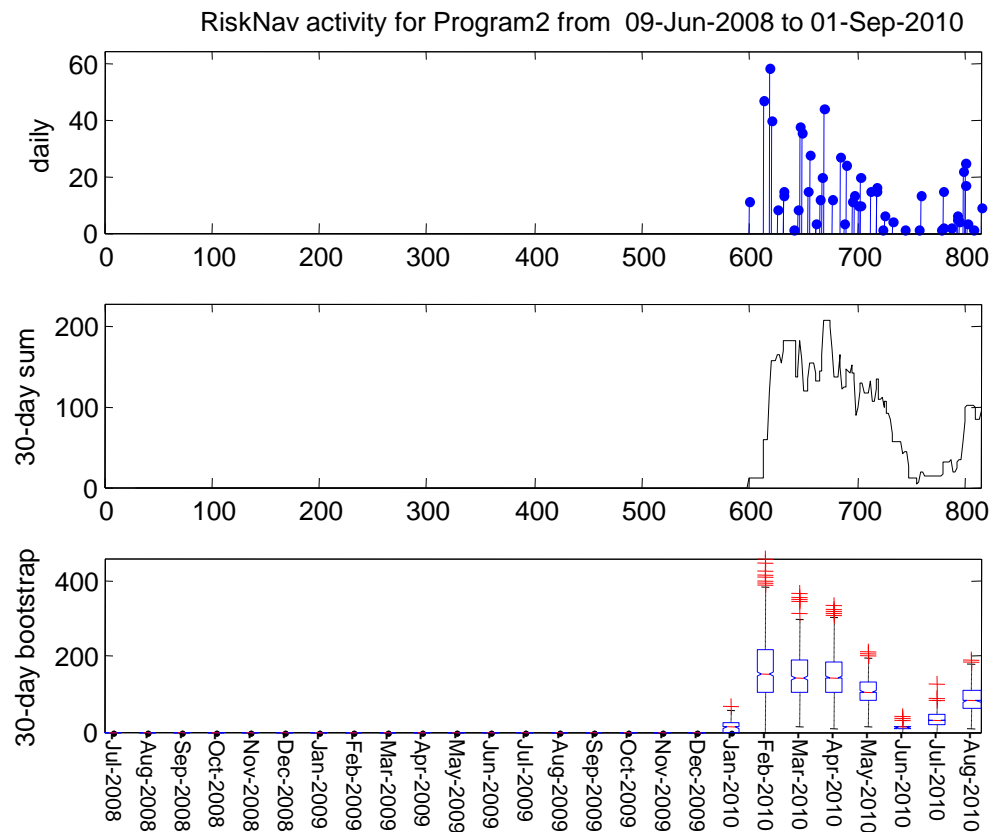
Actionable summary

- Treatment effect concentrated in programs with self-generating risk dialogues
 - *Program2: just coming on-line to RiskNav
 - **Program3: Experienced program cuts in Aug.
- Treatment effect in stimulus driven is observed but discounted
 - Sufficiently large variability in sparse updates

Risk Management		Update	New
Process Maturity	Program	Activity	Risks
Self-Generating	Program1	Y	Y
	Program2	*	Y
	Program3	**	
Stimulus-Driven	Program4	Y	Y
	Program5		
	Program6	Y	
	Program7		
	Program8	Y	Y
Initiating	Program9		
	Program10		Y
	Program11		
Inactive	Program12		
	Program13		
	Program14		

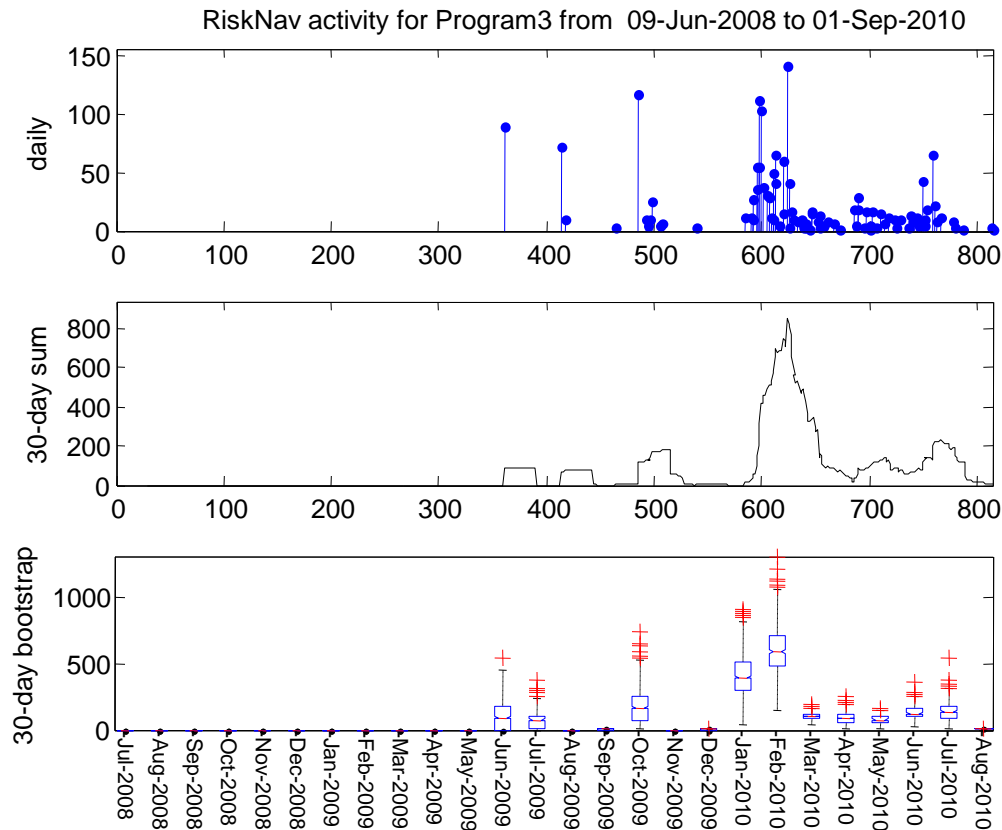
Program2 in more detail

- Program transitioning into RiskNav tool
- Abnormally high activity during baseline period.
- Don't see a treatment effect during test period in update activity
- Do see a treatment effect for new risks



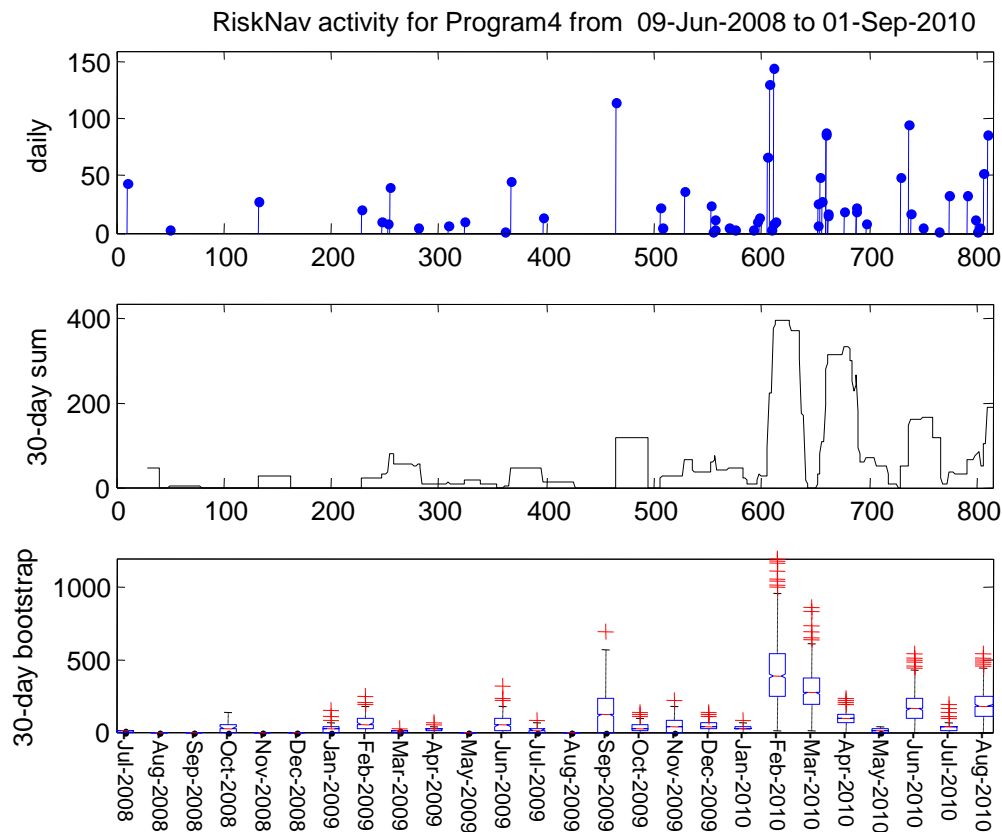
Program3 in more detail

- Mission fundamentally re-evaluated during test period
 - Program cuts August 2010
- Treatment effect is observed comparing {July} vs. {April-May}

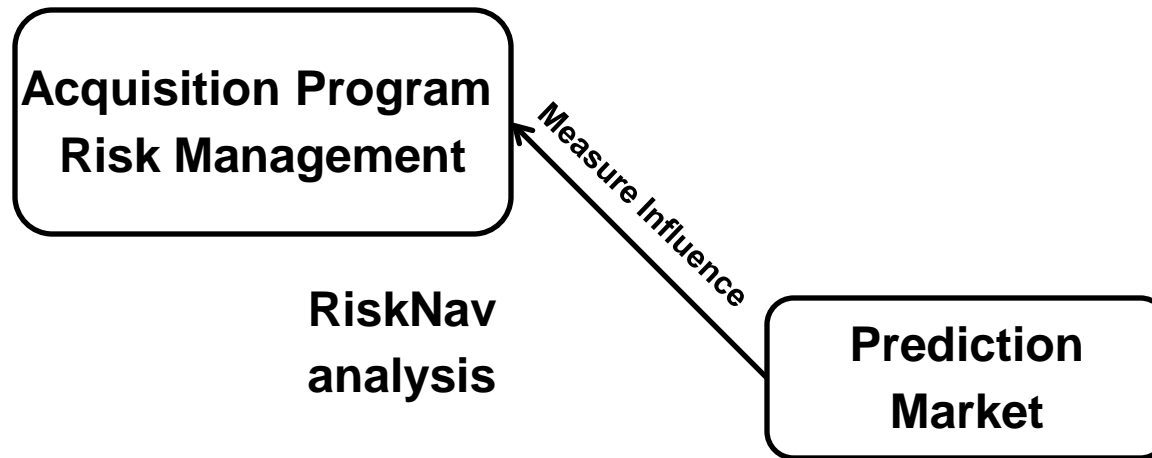


Program4 in more detail

- 100+ update day occurred in June
- Had that stimulus-day occurred in May or July, would have come to different conclusions
- Need longer observation time to declare/deny treatment effect



'Treatment Effects' we would expect to see



- Is there an increase in the overall database activity?
- Is there an increase the rate of newly identified risks?
- Are new risks identified earlier from their event-horizon?
- Do risks get mitigated or closed more quickly?

Risk lead time

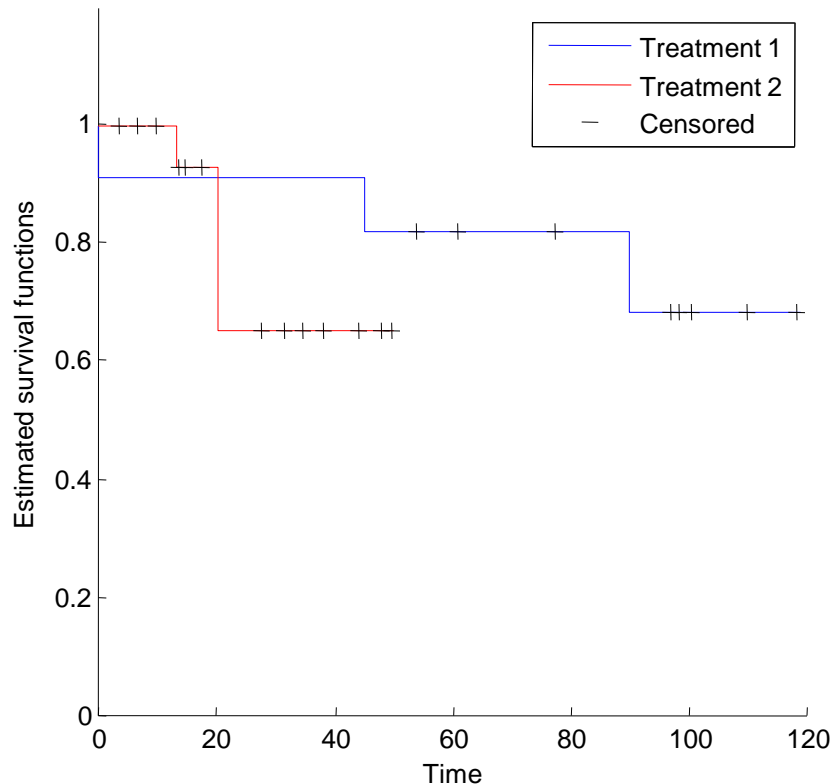
- Subset of new risks specified a lead time:
- Wilcoxon rank-sum test: Do two populations have equally large values?
 - Insufficient data to draw any conclusion

Program	#newRisks (base)	#newRisks (test)	medianImpactDays (base)	medianImpactDays (test)	Wilcoxon p-value
Program1	10	13	109	82	0.120853576
Program12	0	0			
Program2	3	1	476	42	0.5
Program9	0	0			
Program10	0	3			
Program5	1	1	24	101	1
Program6	0	0			
Program11	3	6	112	175	0.547619048
Program4	0	7			
Program3	1	0			
Program13	0	0			
Program7	2	1	161	95	1
Program14	0	0			
Program8	0	4			

Risk closures

- Constructed Kaplan-Mier survival curves for new risks
- Log-Rank test determines if one population survives longer than another
 - Insufficient data to declare closure times are shorter

LogRank test for difference in Kaplan-Meier survival function for Program1 between base and test period



Possible concerns with this analysis

■ Seasonal adjustments

- Insufficient RiskNav histories to make meaningful adjustments
- Traditionally, July-August is a period of reduced work activity
 - We observe an increase in risk-management activity despite this

■ Connection with the Hawthorne effect

- This analysis doesn't focus on specific questions or derived information
- Improvements may be result of general employee feedback in a prediction market

■ This work was exploratory research, not a randomized clinical trial

- We cannot statistically rule out some other (non-prediction market) effect going on here

■ However

- This provides strong direction for future research inquiries
- Analogous to passing 'Phase-0' trials in FDA approval process

Conclusions

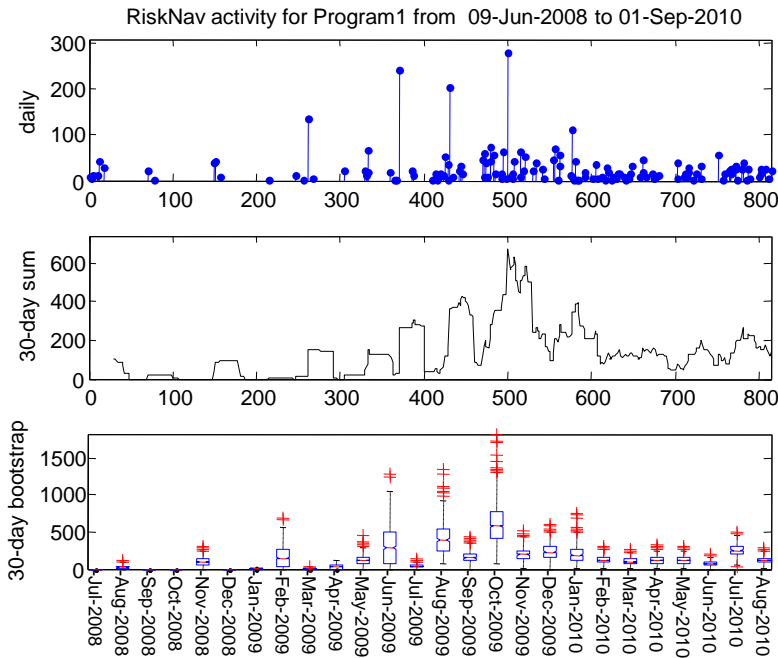
- **After introducing a prediction market to the USAF...**
 - **We observe enhanced risk management practices on programs with self-generating risk identification and mitigation activity**
 - **We do not conclusively observe this effect on programs with stimulus-driven risk management processes**
- **There was not enough data to evaluate whether prediction markets would foster earlier risk identification or faster closure of risks**

Update Activity and New Risks

Program Analysis Data

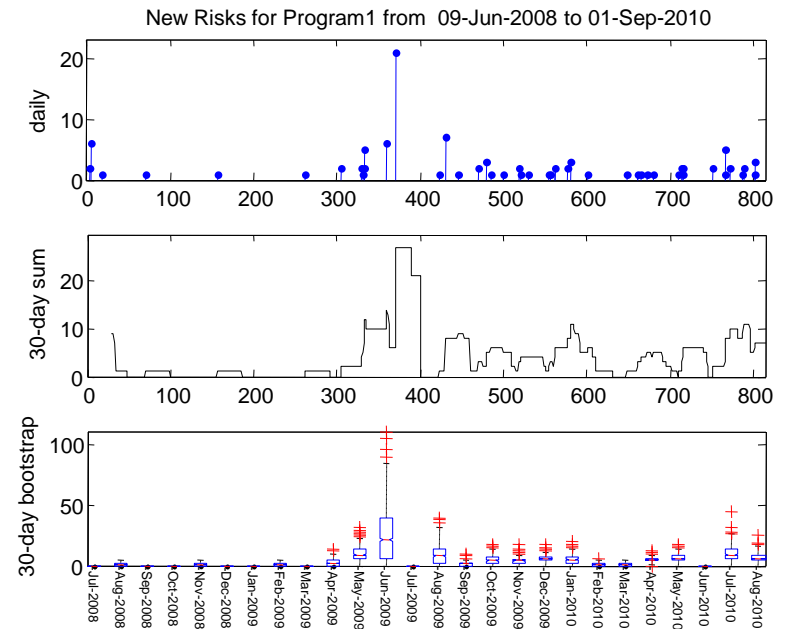
Program1

Update Activity



# Observed Base	246
# Observed Test	381
# at .05 Level	376
p-value	.044

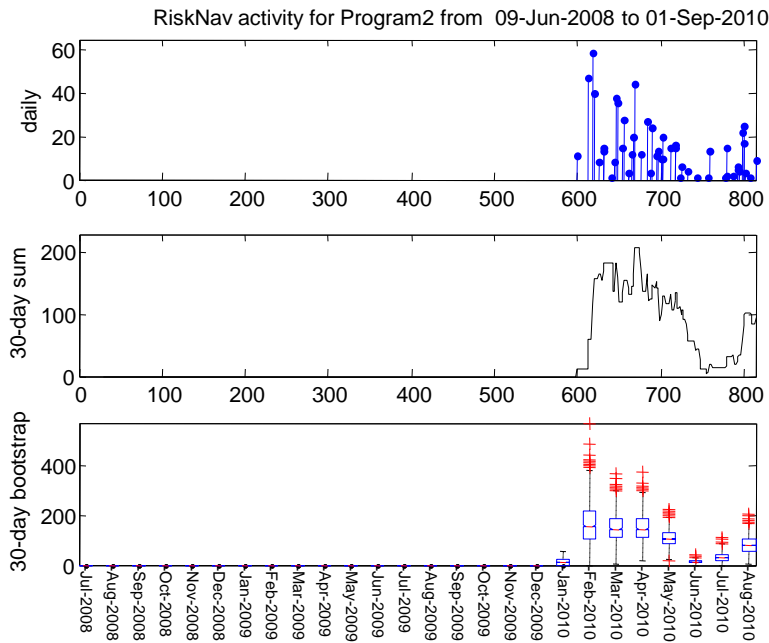
New Risks Identified



# Observed Base	11
# Observed Test	17
# at .05 Level	17
p-value	.047

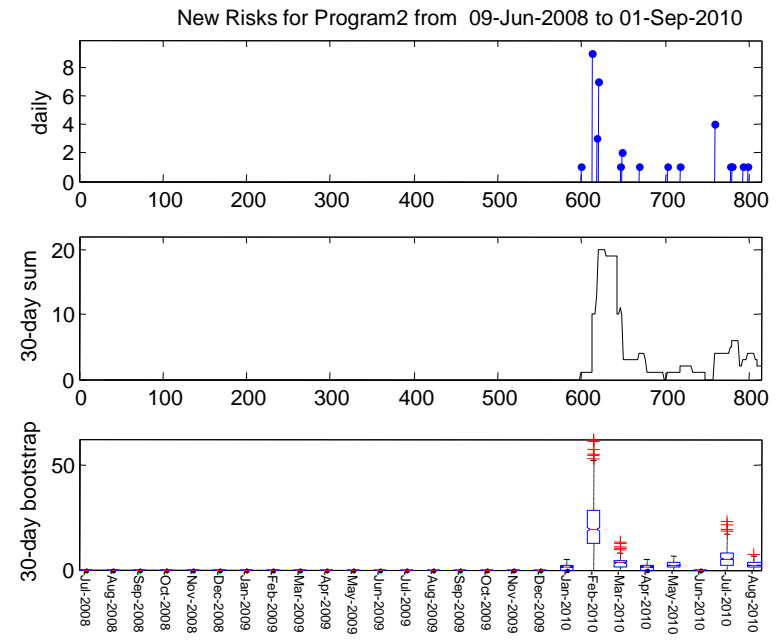
Program2

Update Activity



# Observed Base	258
# Observed Test	177
# at .05 Level	377
p-value	.991

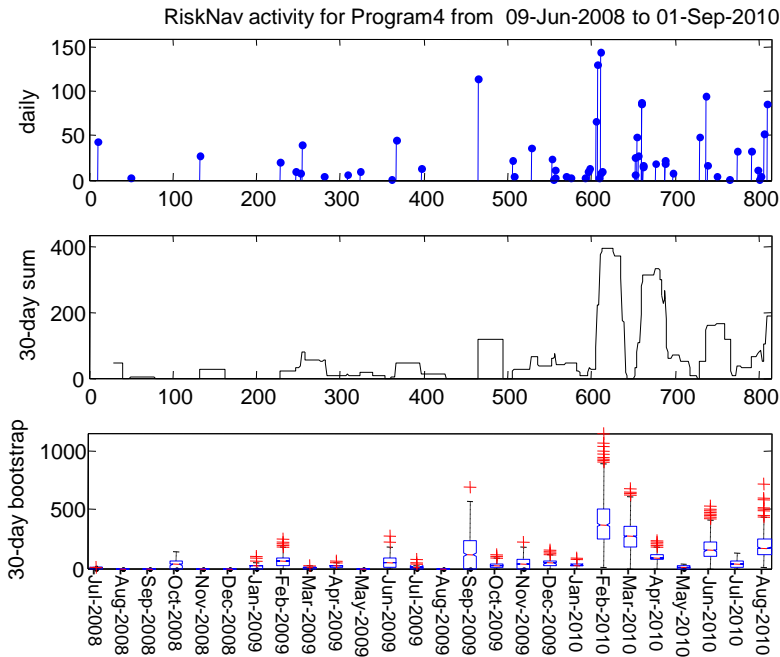
New Risks Identified



# Observed Base	3
# Observed Test	8
# at .05 Level	6
p-value	.003

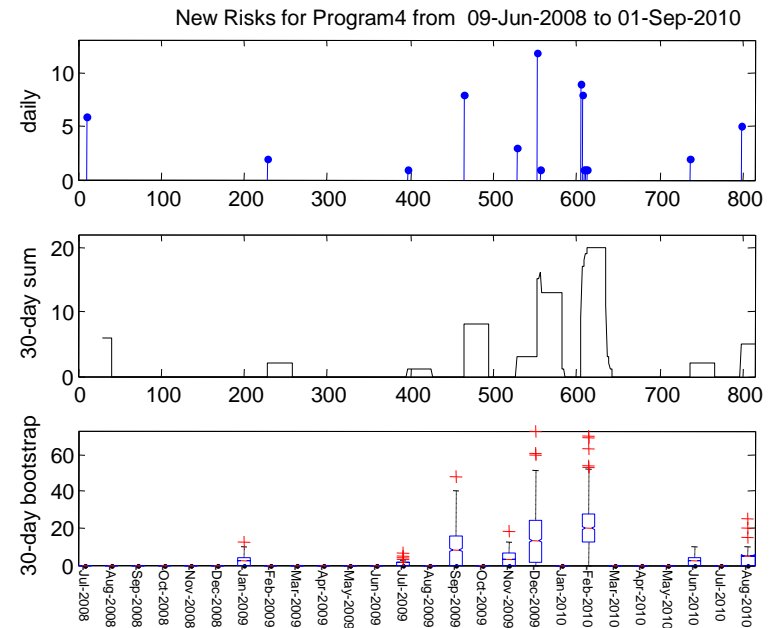
Program4

Update Activity



# Observed Base	100
# Observed Test	222
# at .05 Level	173
p-value	.003

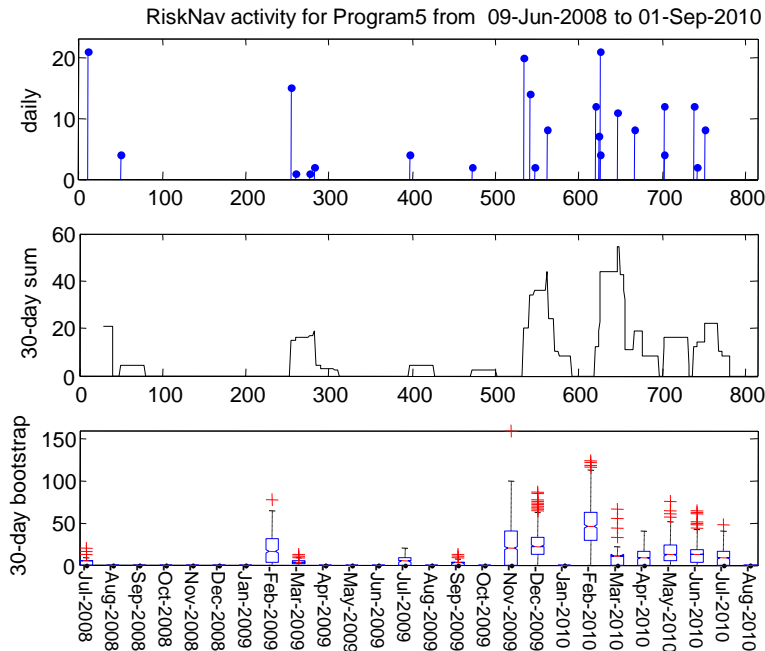
New Risks Identified



# Observed Base	0
# Observed Test	5
# at .05 Level	0
p-value	.000

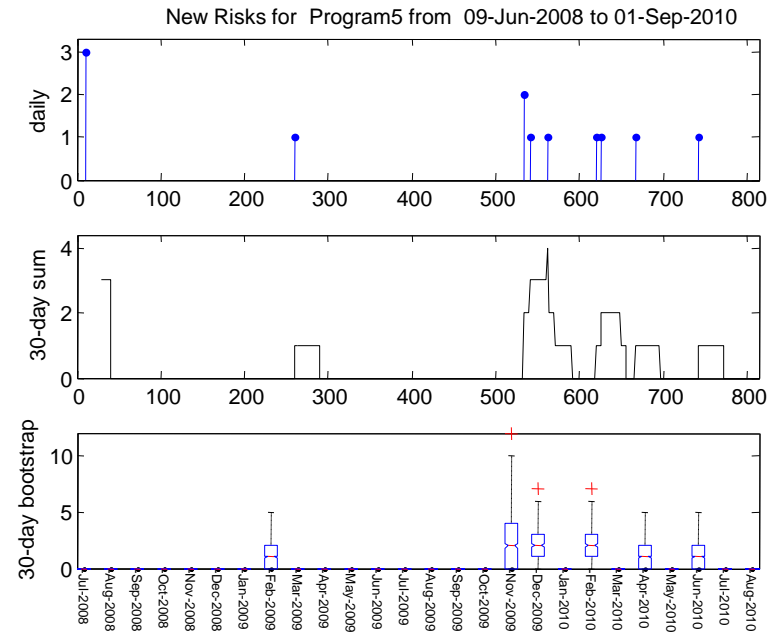
Program5

Update Activity



# Observed Base	24
# Observed Test	8
# at .05 Level	52
p-value	.839

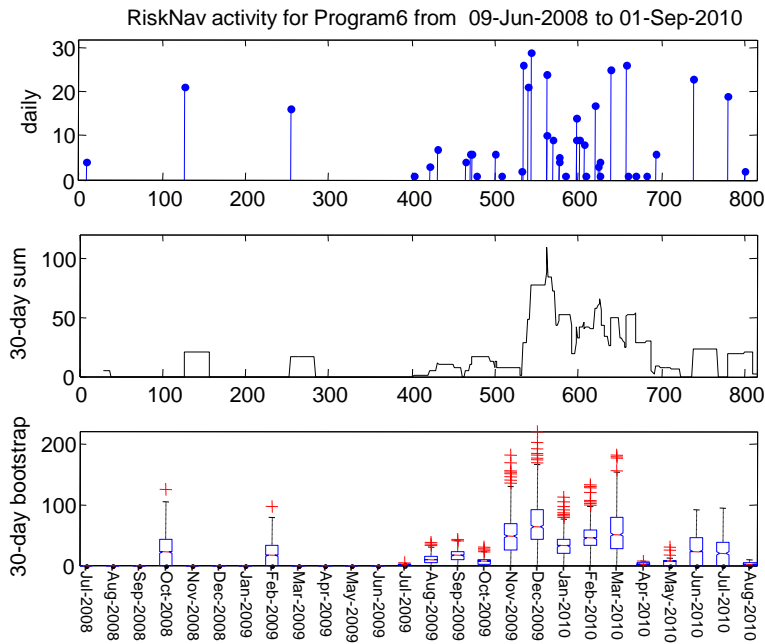
New Risks Identified



# Observed Base	1
# Observed Test	0
# at .05 Level	3
p-value	..636

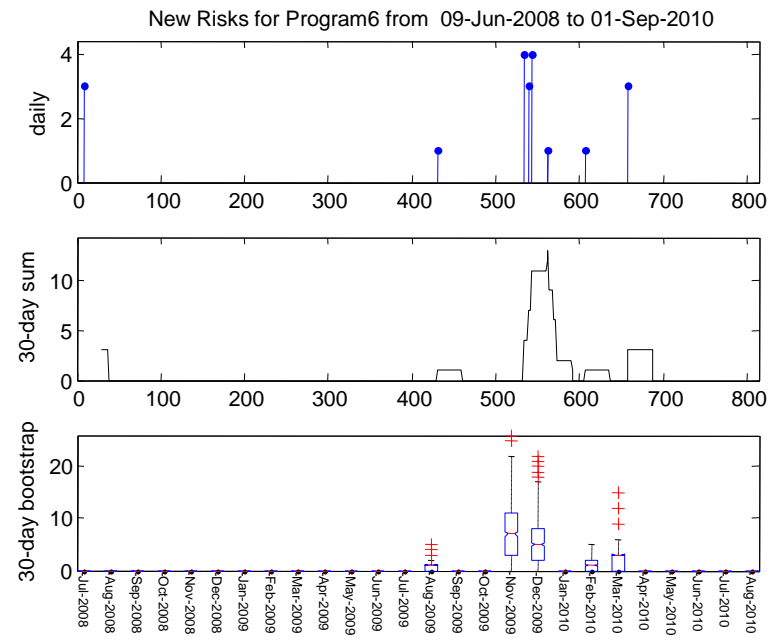
Program6

Update Activity



# Observed Base	8
# Observed Test	21
# at .05 Level	20
p-value	.027

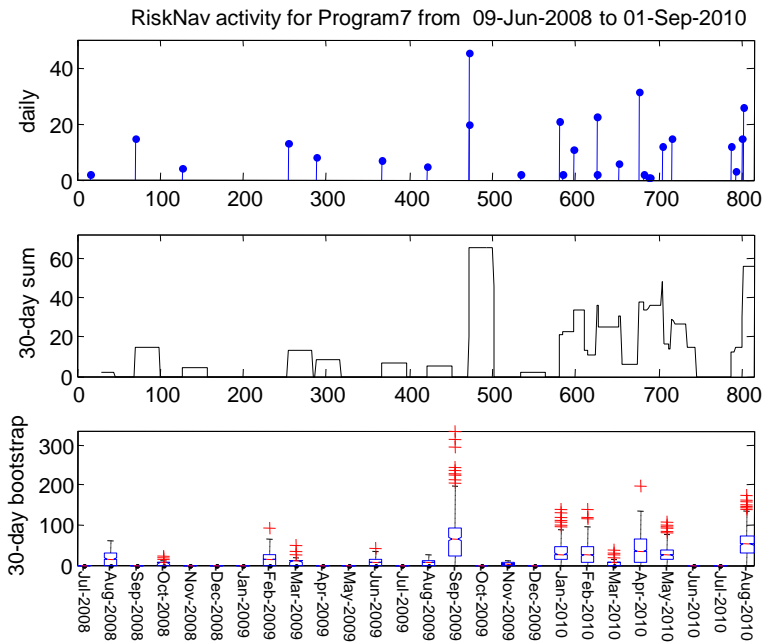
New Risks Identified



# Observed Base	0
# Observed Test	0
# at .05 Level	0
p-value	

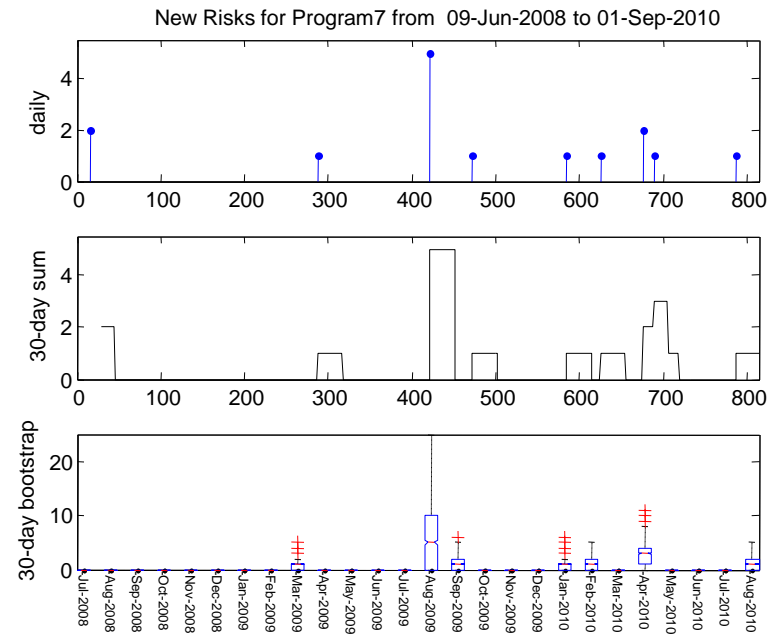
Program7

Update Activity



# Observed Base	63
# Observed Test	56
# at .05 Level	132
p-value	.541

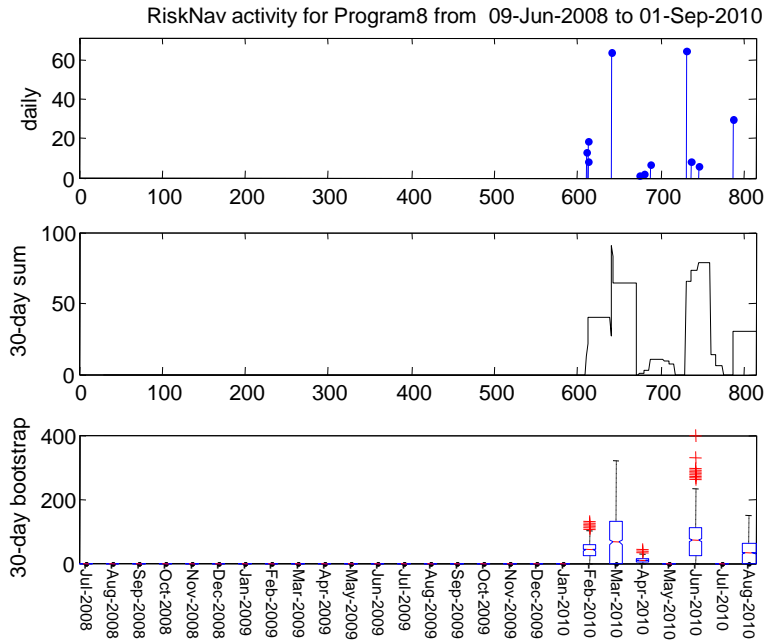
New Risks Identified



# Observed Base	3
# Observed Test	1
# at .05 Level	7
p-value	.747

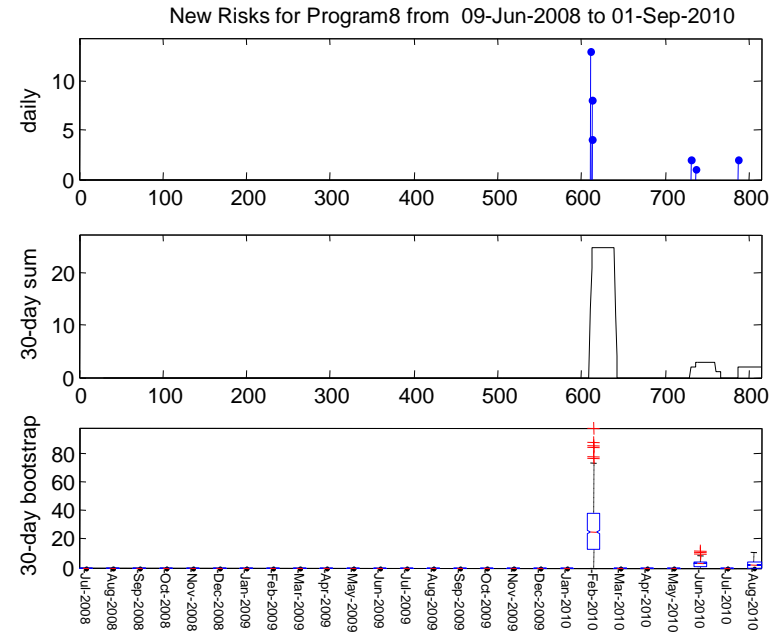
Program8

Update Activity



# Observed Base	10
# Observed Test	30
# at .05 Level	24
p-value	.012

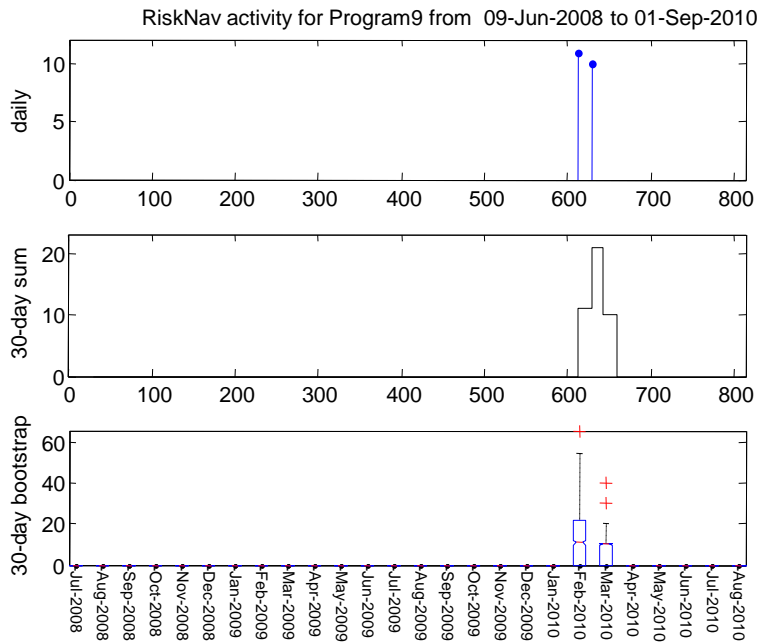
New Risks Identified



# Observed Base	0
# Observed Test	2
# at .05 Level	0
p-value	.000

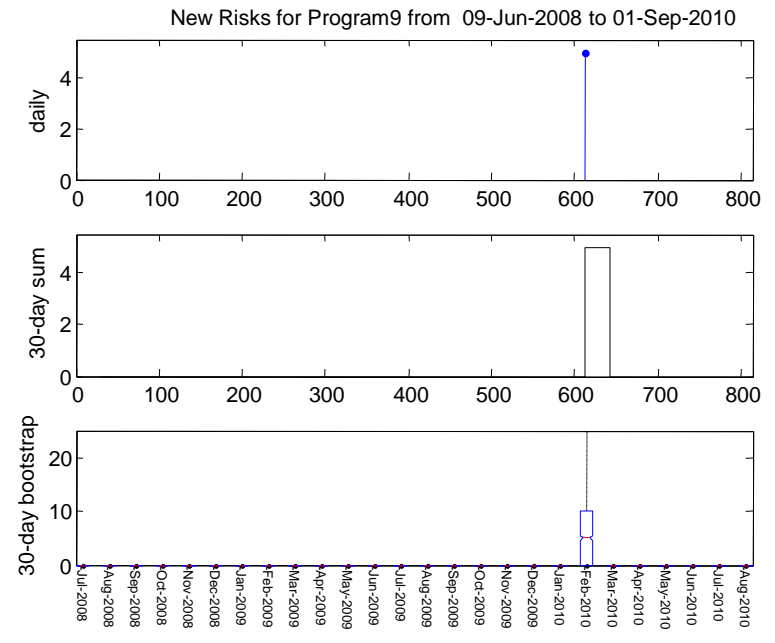
Program9

Update Activity



# Observed Base	0
# Observed Test	0
# at .05 Level	0
p-value	

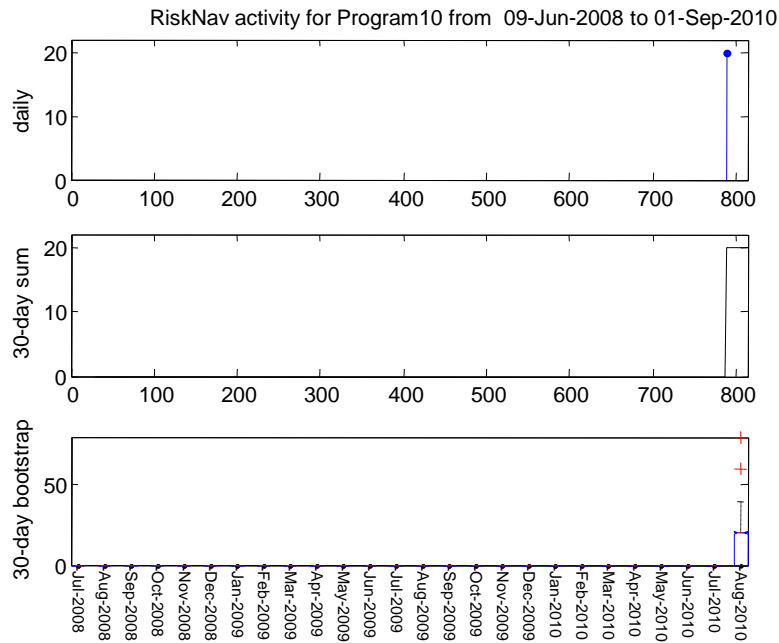
New Risks Identified



# Observed Base	0
# Observed Test	0
# at .05 Level	0
p-value	

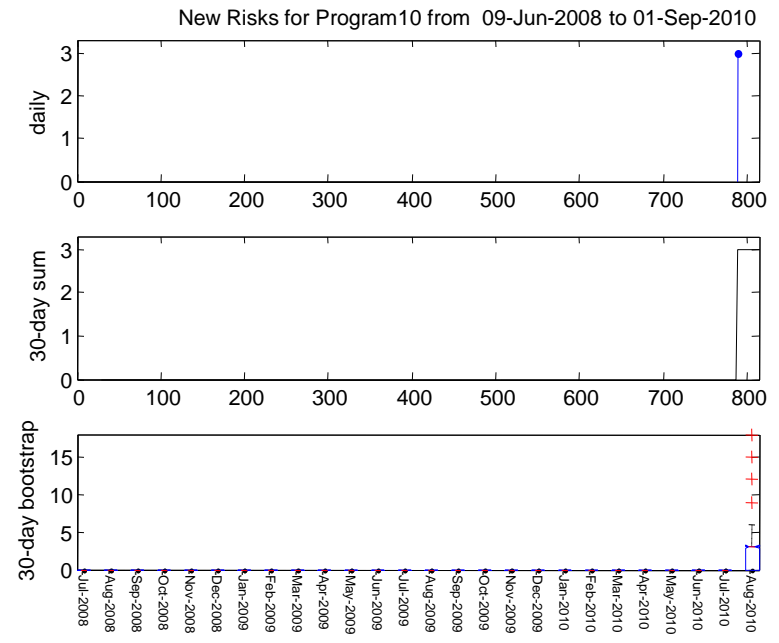
Program10

Update Activity



# Observed Base	0
# Observed Test	20
# at .05 Level	0
p-value	.000

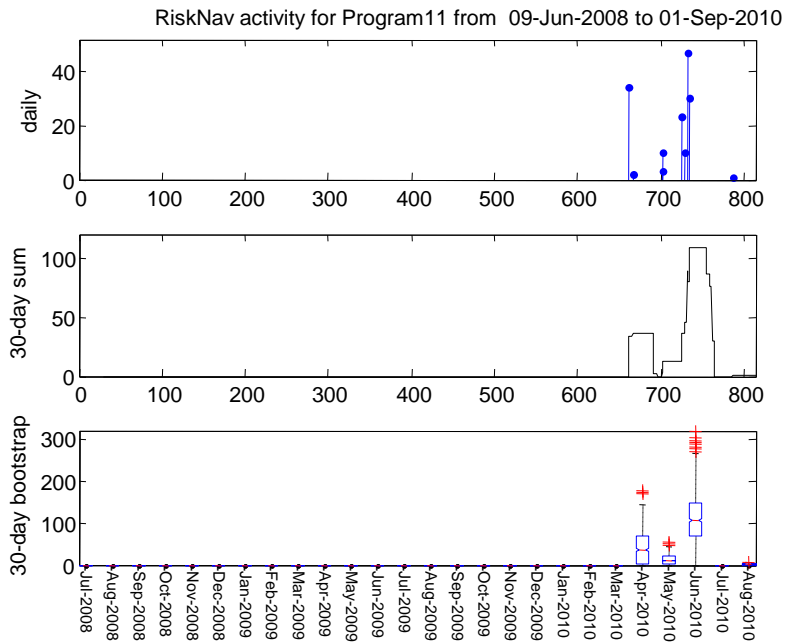
New Risks Identified



# Observed Base	0
# Observed Test	3
# at .05 Level	0
p-value	.000

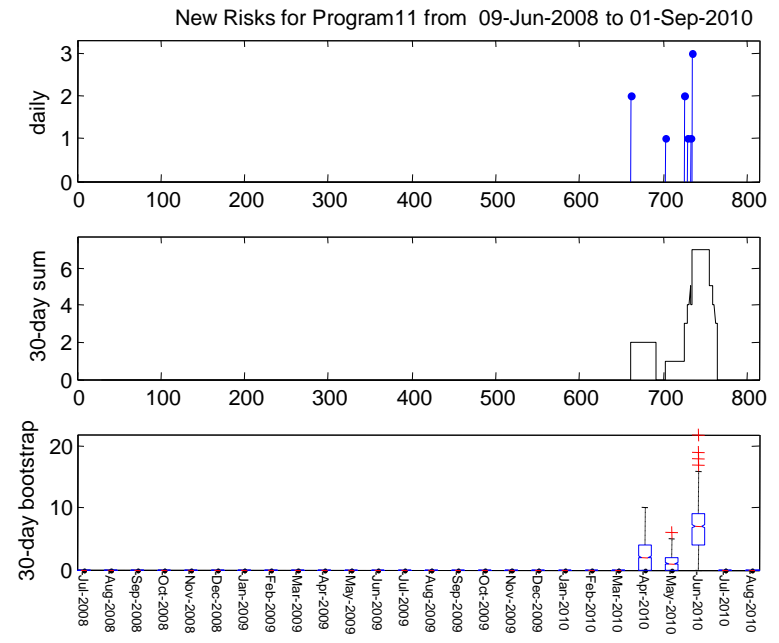
Program11

Update Activity



# Observed Base	49
# Observed Test	1
# at .05 Level	117
p-value	.967

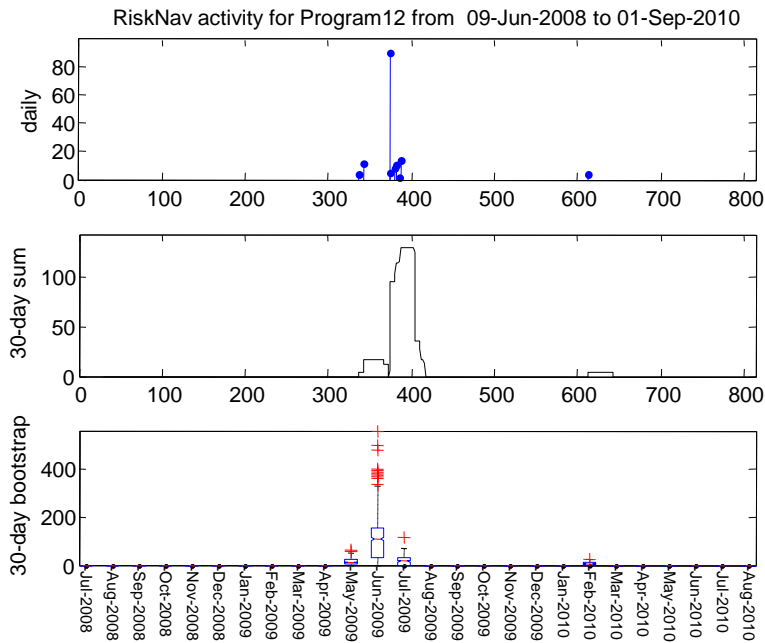
New Risks Identified



# Observed Base	3
# Observed Test	0
# at .05 Level	7
p-value	.870

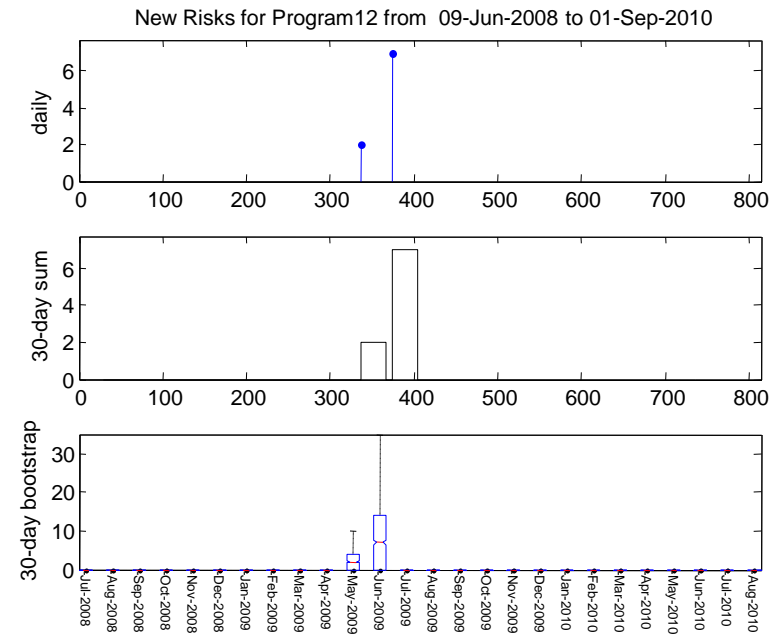
Program12

Update Activity



# Observed Base	0
# Observed Test	0
# at .05 Level	0
p-value	

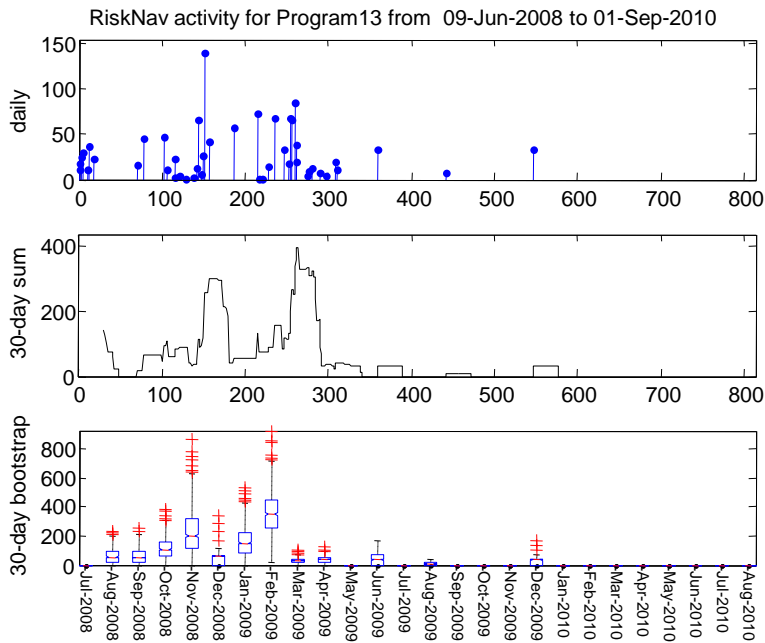
New Risks Identified



# Observed Base	0
# Observed Test	0
# at .05 Level	0
p-value	

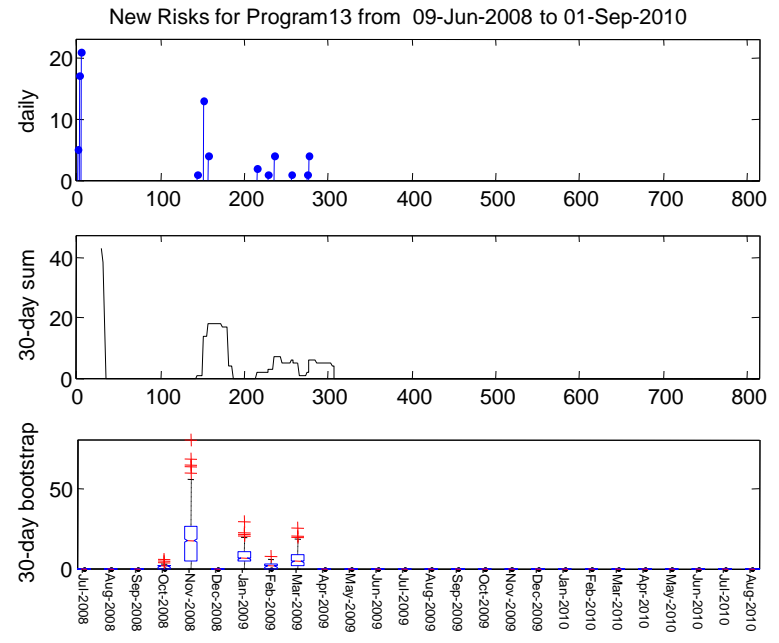
Program13

Update Activity



# Observed Base	0
# Observed Test	0
# at .05 Level	0
p-value	

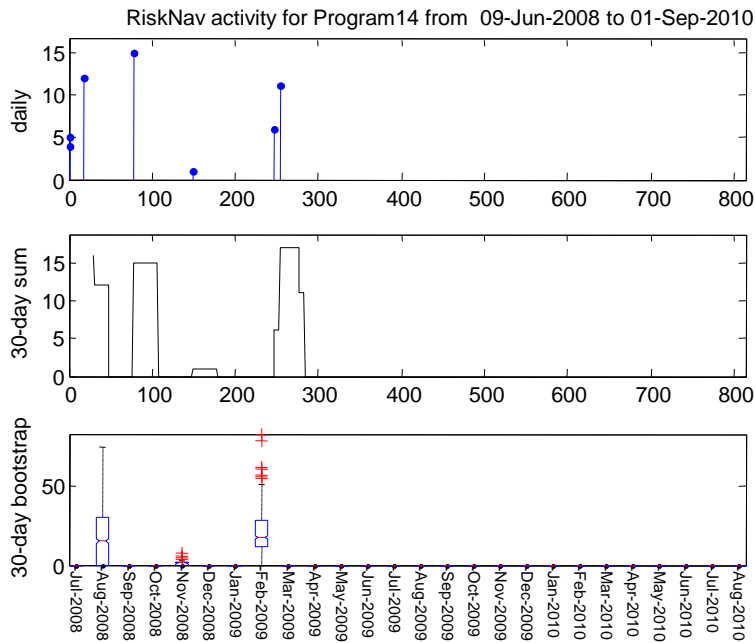
New Risks Identified



# Observed Base	0
# Observed Test	0
# at .05 Level	0
p-value	

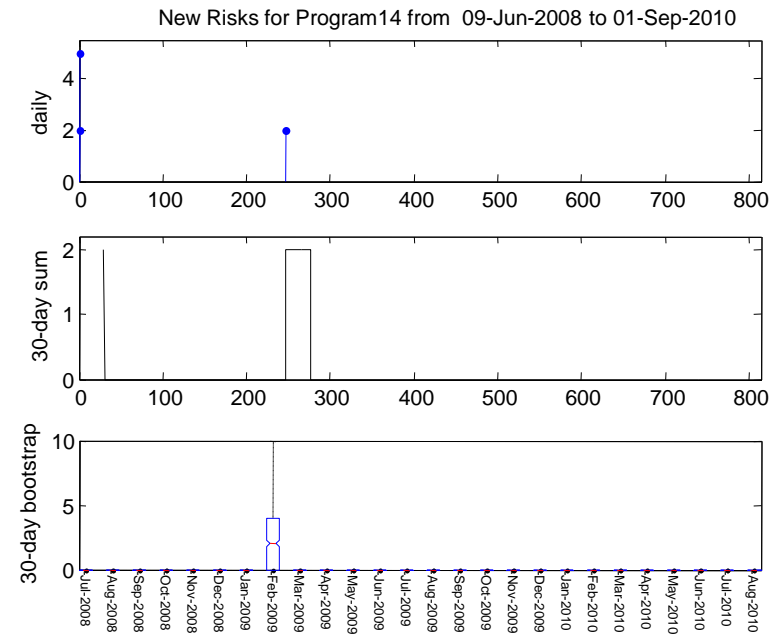
Program14

Update Activity



# Observed Base	0
# Observed Test	0
# at .05 Level	0
p-value	

New Risks Identified



# Observed Base	0
# Observed Test	0
# at .05 Level	0
p-value	