Assessing the Offensive Productivity of NHL Players Using In-game Win Probabilities

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Imagine that you're an NHL GM. You want to add some goal scoring talent to your roster.





Two potential acquisitions





24 years old 15 goals in 2014-2015 0.26 goals per game (career) 48.1% Corsi-tied (career) 24 years old 16 goals in 2014-2015 0.30 goals per game (career) 47.6% Corsi-tied (career)





Which player should you trade for? The one who scores in high pressure situations.

or

The one who is more clutch





Outline of this talk

An NHL Win Probability Model

Narrative Building using Win Probabilities

Measuring Offensive Productivity with Added Goal Value





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ticketmaster*



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An NHL Win Probability Metric

 Hockey journalists and statisticians currently lack many empirical tools available in other sports

One such tool is a metric for calculating second-by-second win probabilities

- ▶ The metric introduced here:
 - ▶ is the only such metric currently available
 - ▶ incorporates powerplay information in a novel way
 - ▶ provides a flexible framework which can unify the work being done on puck possession, zone starts, spatial data, etc.





$$P_{t}(w) = P_{t}(w|\delta_{t}+1) \cdot \Lambda(\gamma_{h} \cdot \nu_{t}) + P_{t}(w|\delta_{t}-1) \cdot \Lambda(\gamma_{a} \cdot \nu_{t}) + P_{t}(w|\delta_{t})(1 - \Lambda(\gamma_{h} \cdot \nu_{t}))(1 - \Lambda(\gamma_{a} \cdot \nu_{t}))$$

When the teams are playing at full-strength, the vector ν_t is all zeros. This means that the metric simplifies to:

$$P_t(w) = P_t(w|\delta_t)$$





$$P_t(w) = P_t(w|\delta_t + 1) \cdot \Lambda(\gamma_h \cdot \nu_t) + P_t(w|\delta_t - 1) \cdot \Lambda(\gamma_a \cdot \nu_t) + P_t(w|\delta_t)(1 - \Lambda(\gamma_h \cdot \nu_t))(1 - \Lambda(\gamma_a \cdot \nu_t))$$

If the home team is on a powerplay then:





$$P_t(w) = P_t(w|\delta_t + 1) \cdot \frac{\Lambda(\gamma_h \cdot \nu_t)}{\Lambda(\gamma_a \cdot \nu_t)} + P_t(w|\delta_t - 1) \cdot \Lambda(\gamma_a \cdot \nu_t) + P_t(w|\delta_t)(1 - \Lambda(\gamma_h \cdot \nu_t))(1 - \Lambda(\gamma_a \cdot \nu_t))$$

If the home team is on a powerplay then:

• $\Lambda(\gamma_h \cdot \nu_t)$ is the probability that the home team scores a PP goal





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If the home team is on a powerplay then:

- $\Lambda(\gamma_h \cdot \nu_t)$ is the probability that the home team scores a PP goal
- $\Lambda(\gamma_a \cdot \nu_t)$ is the probability that the away team scores a SH goal





$$P_t(w) = P_t(w|\delta_t + 1) \cdot \Lambda(\gamma_h \cdot \nu_t) + P_t(w|\delta_t - 1) \cdot \Lambda(\gamma_a \cdot \nu_t) + P_t(w|\delta_t)(1 - \Lambda(\gamma_h \cdot \nu_t))(1 - \Lambda(\gamma_a \cdot \nu_t))$$

If the home team is on a powerplay then:

- $\Lambda(\gamma_h \cdot \nu_t)$ is the probability that the home team scores a PP goal
- ▶ $\Lambda(\gamma_a \cdot \nu_t)$ is the probability that the away team scores a SH goal
- $(1 \Lambda(\gamma_h \cdot \nu_t))(1 \Lambda(\gamma_a \cdot \nu_t))$ is the probability that neither team scores before they return to even strength





A flexible framework

$$P_t(w) = P_t(w|\delta_t + 1) \cdot \Lambda(\gamma_h \cdot \nu_t) + P_t(w|\delta_t - 1) \cdot \Lambda(\gamma_a \cdot \nu_t) + P_t(w|\delta_t)(1 - \Lambda(\gamma_h \cdot \nu_t))(1 - \Lambda(\gamma_a \cdot \nu_t))$$

Everything in red represents the probability that a goal is scored by one team or the other in some period of time.

The framework here allows for any statistics to be included, as long as you have a model that predicts how the stat impacts goal-scoring rates

- ▶ Which players are on the ice
- ► Zone starts

- ▶ Corsi and Fenwick stats
- Spatial data





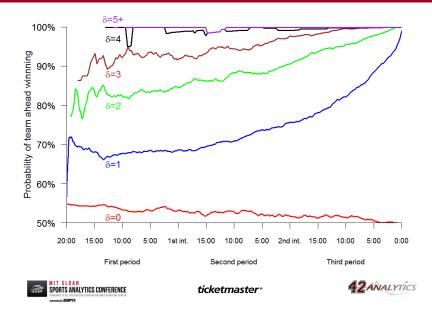
Estimating $P_t(w|\delta_t)$

- I used empirical data to estimate $P_t(w|\delta_t)$
 - \blacktriangleright Data from all regular season games from 2005/2006 through 2012/2013
 - ► Over 9000 games
- I then used a weak Bayesian prior to smooth out the data

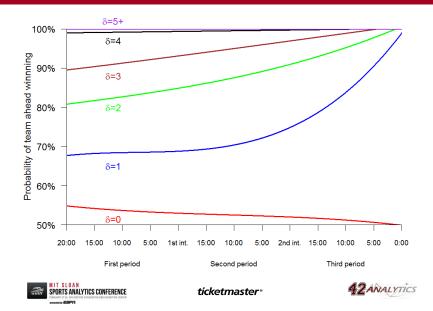




Estimating $P_t(w|\delta_t)$



Estimating $P_t(w|\delta_t)$



Outline of this talk

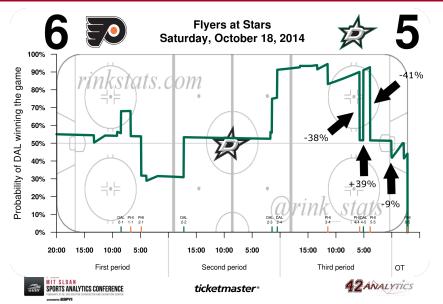
An NHL Win Probability Model

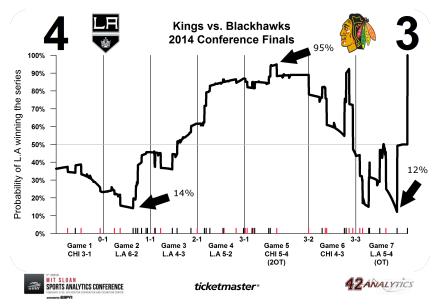
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Win probability apps



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Added Goal Value

Not all goals are created equal.

A 6^{th} goal in a 5-0 game is less valuable to a team than an overtime winner.

Traditional statistics like total goals masks this fact and treats every goal as equally important.

Added Goal Value (AGV) uses win probabilities to measure the significance of each goal. Players with high AGV can be thought of as being more "clutch."





The math behind Added Goal Value

$$AGV_{Crosby} = \sum_{k=1}^{K} \Delta_k - \frac{\sum_{j=1}^{J} \Delta_j}{J} \cdot K$$

$$\Delta_k = P_{t_k+1}(w) - P_{t_k}(w)$$

$$\Delta_j = P_{t_j+1}(w) - P_{t_j}(w)$$

 $\Delta_k = P_{t_k+1}(w) - P_{t_k}(w)$: change in win probability from a goal scored by Sidney Crosby





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 $\Delta_j = P_{t_j+1}(w) - P_{t_j}(w)$: change in win probability from a goal scored another player





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 $\sum_{k=1}^{K} \Delta_k$: total added win percentage for all goals scored by Crosby





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 $\frac{\sum_{k=1}^{K} \Delta_k}{J}: \text{ total added win percentage for all goals scored by Crosby}$ $\frac{\sum_{j=1}^{J} \Delta_j}{J}: \text{ average increase in win probability for all goals (about 17\%)}$





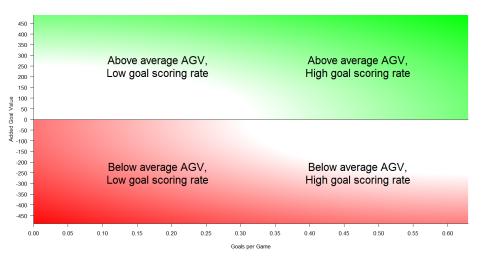
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 $\frac{\sum_{j=1}^{J} \Delta_j}{J} \cdot K:$ expected increase in win probability resulting from the K goals scored by Crosby

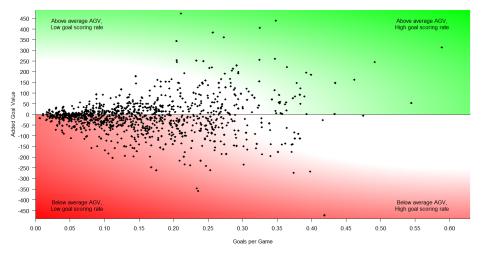






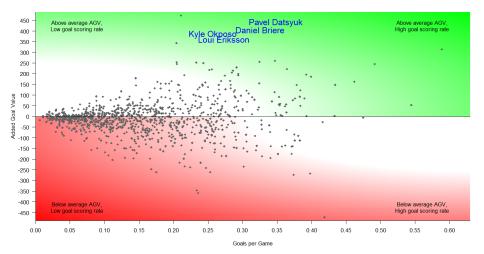






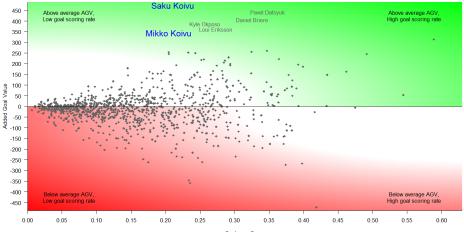






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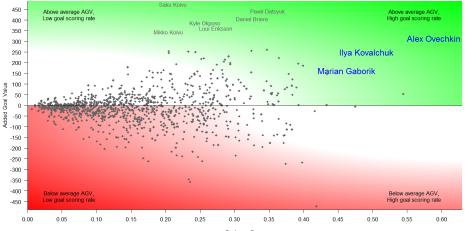




Goals per Game



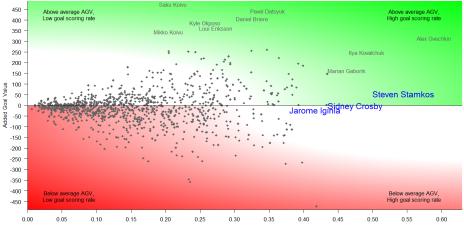




Goals per Game



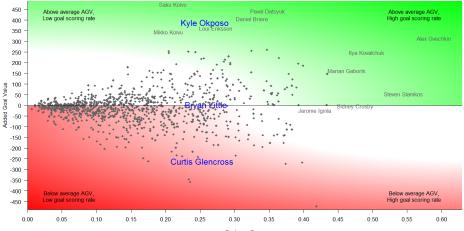




Goals per Game







Goals per Game





Two potential acquisitions



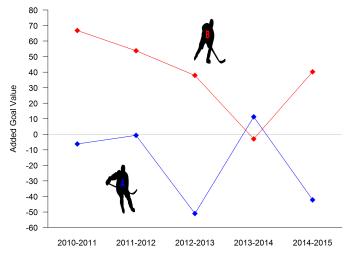


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AGV for Players A and B







Who to acquire?





+40.2 AGV (2014-2015) +198.9 AGV (career)

-42.2 AGV (2014-2015) -88.9 AGV (career)





Next steps

- ▶ Enrich the win probability metric by adding more statistics
- ▶ Stanley Cup winning probabilities; playoff qualification probabilities
- ▶ Adapt AGV to shootout
- ▶ Add assists to AGV metric
- ► AGV app





Thanks for listening!

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