Forecasting Supply and Demand Balance In California's Greenhouse Gas Cap and Trade Market

Elizabeth Bailey, Severin Borenstein, James Bushnell, Frank Wolak and Matt Zaragoza



Cap and Trade: Prices and Quantities

- Economists often frame question of caps vs. taxes as
 - Caps provide emissions certainty
 - Taxes provide price (cost) certainty
- Climate Policy: a more complicated picture
 - Uncapped sectors and regions increase quantity uncertainty
 - Complementary policies cause abatement apart from market price
- California's Cap and Trade is a hybrid
 - Auction reserve price (floor)
 - Price containment reserve ("ceiling?")
- Important to understand how relevant these price collars are
 - Large probability that prices are at either floor into containment reserve



General Approach

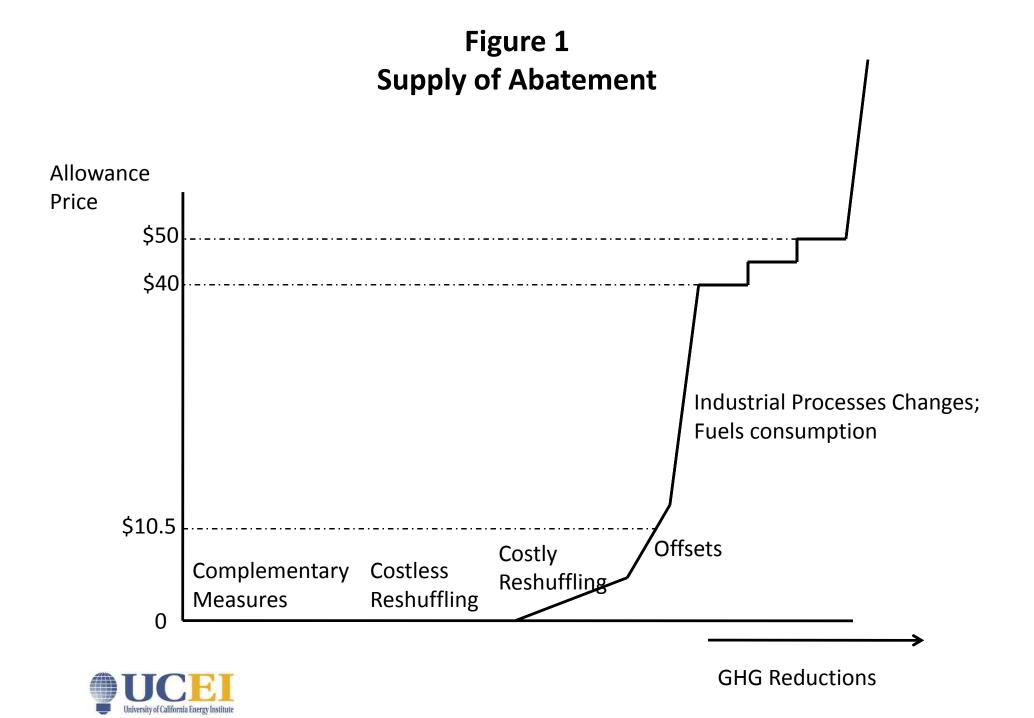
- Estimate probability model for future business-as-usual (BAU) emissions
 - BAU minus cap is "demand" for abatement
- Consider scenarios of complimentary measure impacts
 - Measures that are not directly responsive to allowance prices
- Consider scenarios of abatement ``supply" in response to varying allowance price levels
- Combine these to forecast distribution of future allowance prices
 - Probability of prices at floor
 - Probability of prices in allowance reserve
 - Probability at price above allowance reserve



Assumptions on Timing

- Main analysis assumes years/phases fully integrated over time through banking
- Aggregate emissions, complimentary measures, and abatement over 8 years
- All calculations based upon 8 year totals
- Implication is prices should rise at the rate of interest
 - We report results in 2012 prices
 - Means "real" floor price is about \$15
 - Breakdown in banking increases risks of price swings





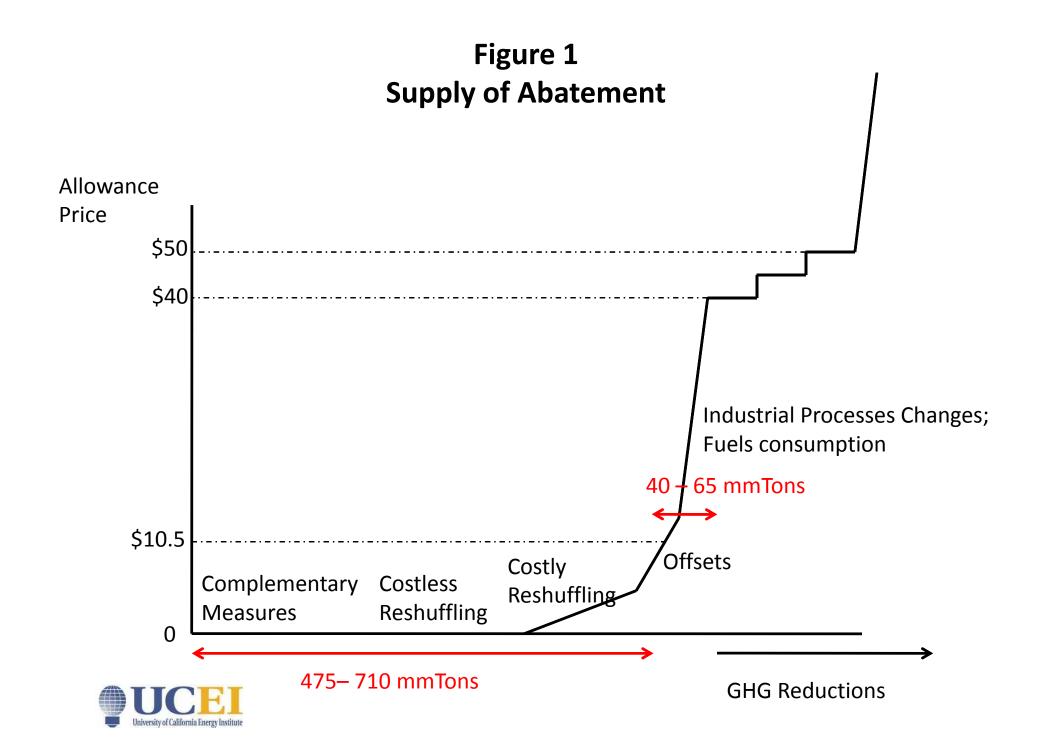


Figure 2
Hypothetical Distribution of Abatement Demand (BAU minus Allowances Outside Containment Reserve) vs Abatement Supply

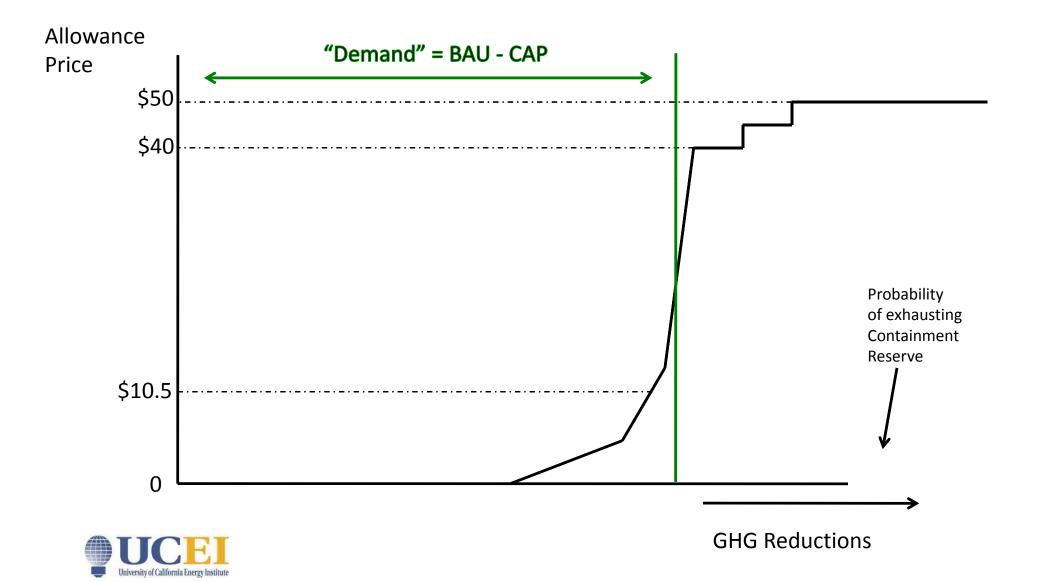


Figure 2
Hypothetical Distribution of Abatement Demand (BAU minus Allowances Outside Containment Reserve) vs Abatement Supply

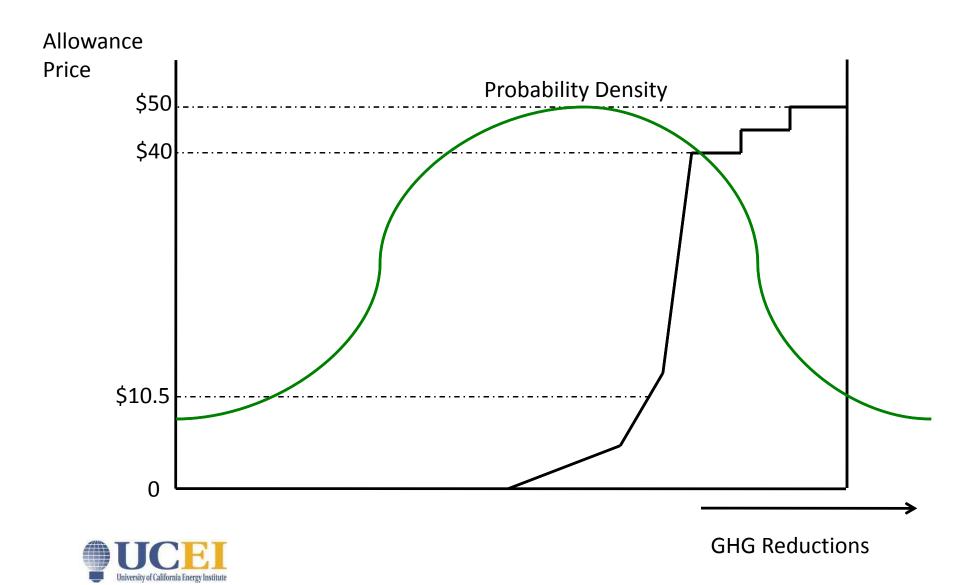


Figure 2
Hypothetical Distribution of Abatement Demand (BAU minus Allowances Outside Containment Reserve) vs Abatement Supply

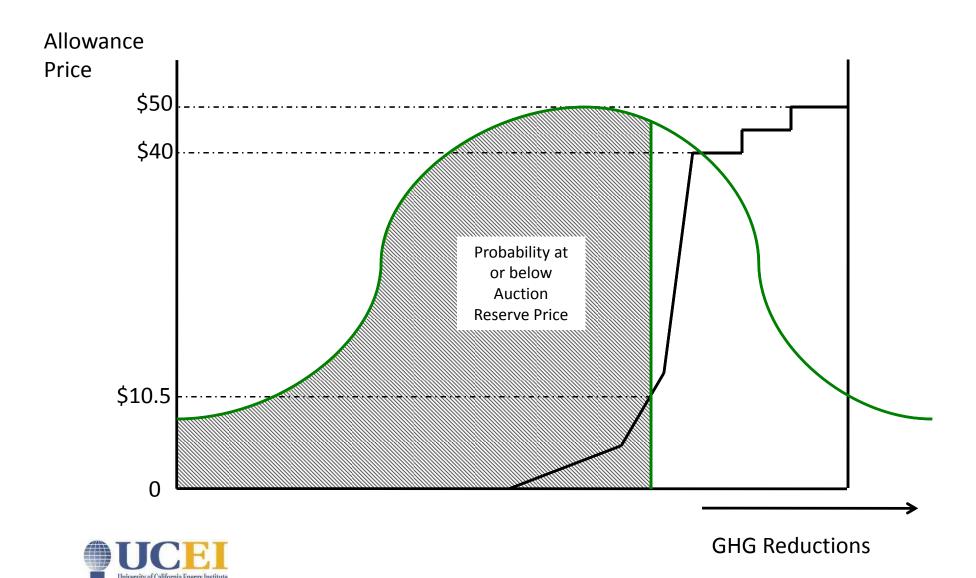
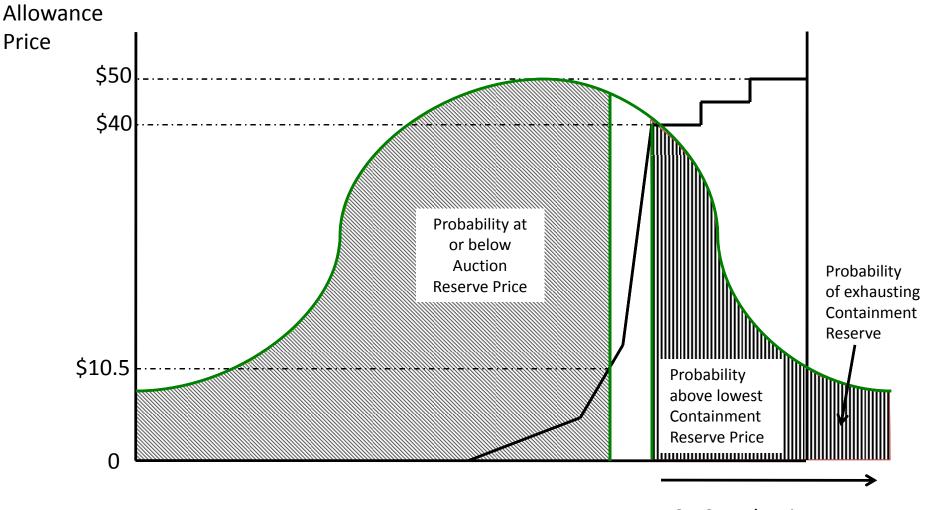


Figure 2
Hypothetical Distribution of Abatement Demand (BAU minus Allowances Outside Containment Reserve) vs Abatement Supply





GHG Reductions

Figure 2
Hypothetical Distribution of Abatement Demand (BAU minus Allowances Outside Containment Reserve) vs Abatement Supply

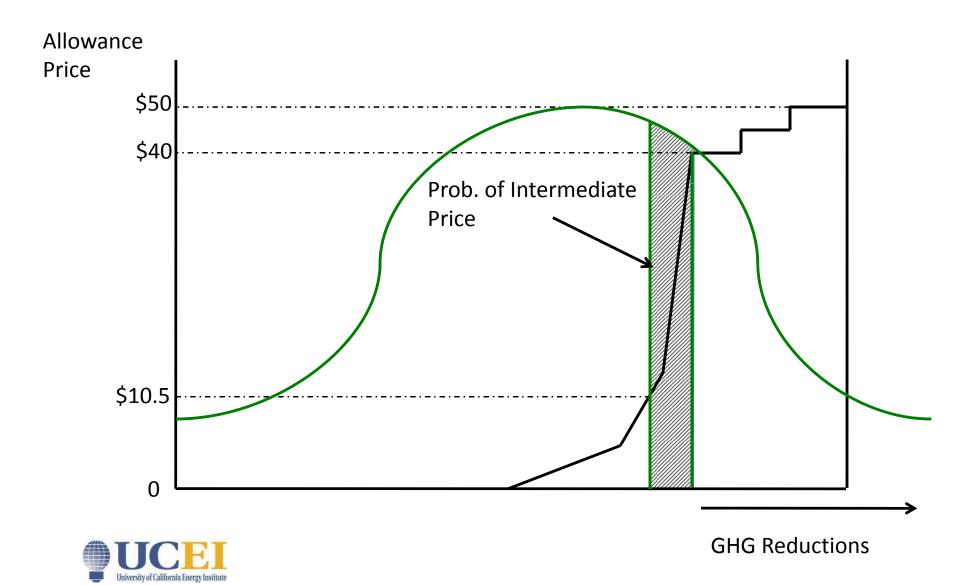
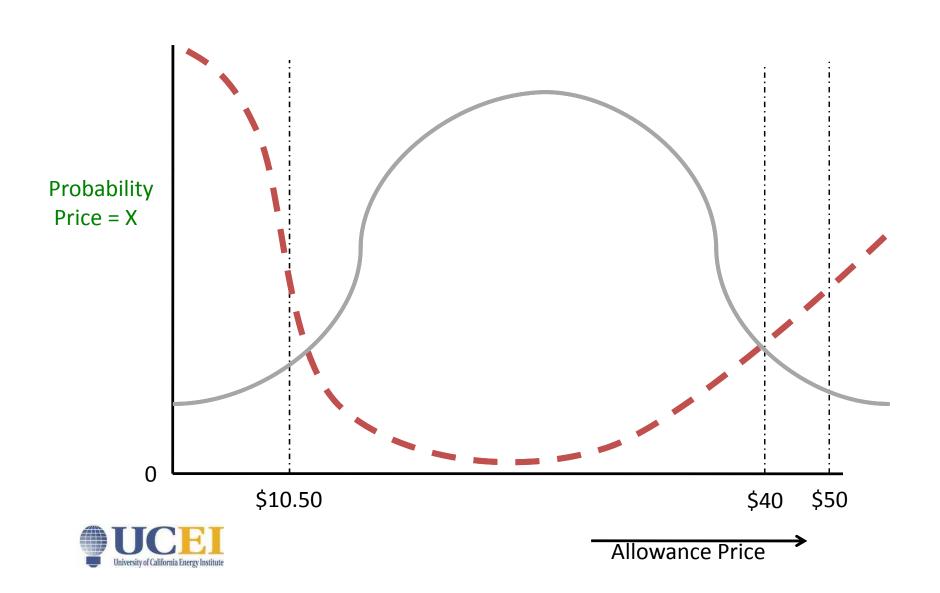


Figure 3
Possible Density Functions of Allowance Price



Model for BAU Emissions

- Estimate time series model of drivers of sectoral greenhouse gas (GHG) emissions and sectoral emissions intensity
 - Includes Industrial & Nat Gas emissions, vehicle emissions, electricity emissions, GSP, VMT, Electricity consumption, Oil price
 - Model allows for economic activity in sector to increase yet GHG intensity of sector to fall
- Use model to construct a distribution of future GHG emissions from 2013 to 2020 that accounts for
 - Uncertainty in econometric model parameter estimates
 - Estimation Error
 - Uncertainty in future values of unobservables in econometric model
 - Prediction Error
- Model assumes all variables are 2nd-order stationary in growth rates but allows for linear combinations of elements of model to be stationary
 - Co-integration restrictions imposed in estimation and simulation that reflect "equilibrium" relationships between variables in model
 - Imposing these restrictions improves forecasting accuracy of model
 - Model uncertainty is a third source of uncertainty
 - Not explicitly taken into account in distribution of future GHG emissions, but estimated distributions are very similar across a wide range of models that assumed variables are 2nd-order stationary in growth rates

Model for BAU Emissions

From F[$(Y_{T+1}, Y_{T+2},, Y_{T+9}, Y_{T+10})$ |GSP(2011),GSP(2012)] can apply change of variables to compute an estimate of distribution of state-wide GHG emissions for Phase I = electricity and industrial processes for 2013, 2014 Phase II = Phase I + transportation and natural gas for 2015 to 2020

Cap three emissions intensity measures at in-sample median, 75th percentile, or maximum in constructing future GHG emissions

If realized value of intensity is greater than cap, then re-set value to equal cap and multiply by economic activity measure to obtain sectoral GHG emissions

Compute distribution of cumulative GHG emissions covered by cap for 2013 through 2020

Sum of GHG emissions covered from start of program though end of each year

Report E(Cumulative Sum of Covered GHG Emissions | GSP(2011),GSP(2012)] for each year from 2013 to 2020

Compute pointwise (for each year) upper and lower 95% confidence intervals for each conditional expectation of cumulative annual GHG emissions



Figure 4a Estimated Business-As-Usual Emissions

(with GHG Ratios to Other Factors Bounded Above at Median Levels)

VECM(1) Cumulative CO2 Forecast (kernel density) (conditional on GDP 2011 & 2012, intensities capped at sample median) (model 2) - Mean CI Cumulative CO2 Emission Year



Figure 4b Estimated Business-As-Usual Emissions

(GHG Ratios to Other Factors Bounded Above at 75th Percentile)

VECM(1) Cumulative CO2 Forecast (kernel density) (conditional on GDP 2011 & 2012, intensities capped at sample q3) (model 2) Mean ······CI Cumulative CO2 Emission

Year



Figure 4c Estimated Business-As-Usual Emissions

(GHG Ratios to Other Factors Bounded Above at Maximum)

VECM(1) Cumulative CO2 Forecast (kernel density) (conditional on GDP 2011 & 2012, intensities capped at sample max) (model 2) Mean ---- CI Cumulative CO2 Emission Year



Supply of Abatement

Table 1: Potential Emissions Reductions from Complementary Policies

| Category | Measure | Average Annual | Years Under | Total Aggregate |
|------------------|-------------------|----------------|-------------|------------------|
| | | Reductions | Сар | Reductions |
| | 20% and 33% RPS | 7.8 – 12.4 MMT | 2013 – 2020 | 62.4 – 98.8 MMT |
| Complimentary | Auto Standards | 9.3 – 16.2 MMT | 2015 – 2020 | 74.2 – 129.8 MMT |
| Measures | LCFS | 0 – 10.3 MMT | 2015 – 2020 | 0 – 61.9 MMT |
| | Energy Efficiency | 0 – 3.4 MMT | 2013 – 2020 | 0 – 27 MMT |
| | Other transport | 0 – 1.5 MMT | 2015 – 2020 | 0 – 12.4 MMT |
| Low-price | Offsets | 9.4 – 17.4 MMT | 2013 – 2020 | 75 - 139 MMT |
| Responses | Reshuffling | 15 – 45 MMT | 2013 – 2020 | 120 – 360 MMT |
| | Gasoline | | 2015 - 2020 | 13.4-26.7 MMT |
| Price-Responsive | Natural Gas | | 2015 - 2020 | 18.5-35.8 MMT |
| (at \$50/ton) | Electricity | | 2013 - 2020 | 15-25 MMT |
| Totals | | | | 378.5 – 916.4 |

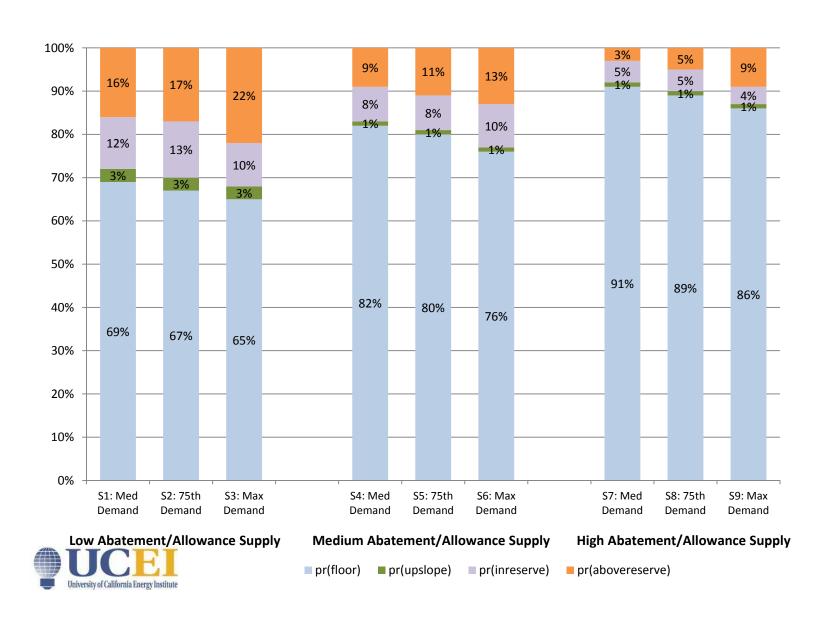


Three Abatement Scenarios

- Low Availability:
 - 475 MMT from comp. and low price policies
 - Medium price response
- Medium Availability:
 - 583 MMT from comp. and low price policies
 - Low price response
- High Availability:
 - 710 MMT from comp. and low price policies
 - Medium price response



Figure 5
Allowance Price Probabilities by Scenario



Conclusions from Modeling

- Very skewed distribution of possible prices
 - "fat" tails on low and high prices
 - Steeper abatement curves and fatter tails on expected emissions magnify this effect
- Expected prices less informative about risks of extreme prices than for normal distributions
- Without banking, ``width'' of the segments from the abatement supply curve narrow
 - Prices would not be the same every year
 - Probability of extreme prices in any given year increase



Policy Implications

- Small, but real chance of reaching and exhausting allowance reserve before 2021.
 - Specific policies to respond to potential exhaustion of reserve are needed. CARB is working on these.
- Allowance revenues could fall well below previous forecasts.
 - Floor price most likely outcome
 - Lower sales at the floor price
- Prices could be volatile as market updates to new information
 - Small swings in BAU or abatement could lead to large prices swings

